

UNIVERSITY OF CINCINNATI

Date: 12-Mar-2010

I, Michael Connelly ,

hereby submit this original work as part of the requirements for the degree of:

Doctor of Philosophy

in Materials Science

It is entitled:

An Analysis of Innovation in Materials

and Energy

Student Signature: Michael Connelly

This work and its defense approved by:

Committee Chair: Jainagesh Sekhar, PhD
Jainagesh Sekhar, PhD

Ronald Huston, PhD
Ronald Huston, PhD

Steven Benintendi, PhD
Steven Benintendi, PhD

Jude Iroh, PhD
Jude Iroh, PhD

Rodney Roseman, PhD
Rodney Roseman, PhD

An Analysis of Innovation in Materials and Energy

A dissertation submitted to the
Graduate School
of the University of Cincinnati
in partial fulfillment of the
requirements for the degree of

Doctor of Philosophy

in the Department of Chemical and Materials Engineering
of the College of Engineering

by

Michael Connelly

M.S. University of Cincinnati

June, 2010

Committee Chair: J.A. Sekhar, Ph.D.

Abstract: This dissertation presents an analysis of innovation in engineering materials and energy sources. More than fifty engineering materials and fourteen energy sources were selected for an evaluation of the relationship between the yearly production activity and yearly patent counts, which may be considered as a measure of innovation, for each. Through the employment of correlation theory, best-fit and origin shift analyses, it has been determined here that engineering materials and energy sources display similar life cycle and innovative activity behaviors. Correlation theory revealed a relationship between the yearly production and yearly patent counts indicating the extent that production and innovation affect each other. Best-fit analysis determined that four-stage life cycles exist for both engineering materials (metals and non-metals) and energy sources. Correlation and best-fit indicators of an estimated Stage III are confirmed by the presence of an origin shift of the patent data when compared to the production data which indicates that patents, or innovation, are driving, or being driven by, production. This driving force could represent the constructive or destructive side of the innovative process, with such sides being delineated by a possible universal constant above which there is destructive innovative behavior and below which exists constructive innovation. The driving force may also illustrate the manner in which an engineering material or energy source transitions into an innovatively less active state, enter Stage IV and possibly become a commodity. A possible Stage V, indicating “Final Death”, is introduced in which production is on a steep decline with no signs of recovery. Additionally, innovatively active energy sources are often found to utilize or be supported by innovatively active engineering materials. A model is presented that can be used for the evaluation of innovation and production that can be applied to both engineering materials and energy sources that may be used to predict the innovative behavior of these resources in order that they can be more effectively allocated and utilized.

Table of Contents

Abstract	Page iii
List of Tables and Figures	Page x
Introduction	Page 1
Section 1 : Measuring Innovation	Page 4
Innovation	Page 4
Measurement of Innovation	Page 5
Measurement Excluding Patents	Page 5
Measurement with Patents	Page 7
Section 2 : Data Collection	Page 9
Engineering Material Production Activity Data Collection .	Page 9
Patent Data Collection	Page 10
Section 3 : Patent and Production Activity Data Correlation	Page 11
Section 4 : Best-Fit	Page 15
Platform Equation with Chromium as an Example	Page 16
Stage Indication	Page 24
Section 5 : Best-Fit, Origin Shift and Innovation	Page 28
Section 6 : Energy Sources	Page 37
Data Collection	Page 37
Patent and Production Activity Data Collection	Page 38
Best-Fit	Page 40
Wind Energy as an Example	Page 41
Best-Fit, Origin Shift and Innovation	Page 45
Energy Materials	Page 49

Engineering Material Connection	Page 51
Result Comparison	Page 53
Section 7 : Analysis	Page 54
Correlation	Page 54
Best-Fit	Page 55
Stage V	Page 56
Origin Shift	Page 57
Driving Force	Page 58
Life Cycle Stage Change	Page 61
Relevance	Page 63
Conclusion and Summary	Page 67
Appendix 1 : Correlation	Page 69
Appendix 2 : MatLab	Page 71
Program Template	Page 71
Zinc Program	Page 72
Appendix 3 : Engineering Materials Data	Page 76
Aluminum	Page 76
Antimony	Page 78
Arsenic	Page 81
Asbestos	Page 83
Barite	Page 85
Bauxite/Alumina	Page 87
Beryllium.....	Page 90
Bismuth	Page 92
Boron	Page 94

Cadmium	Page 96
Chromium	Page 98
Cobalt	Page 100
Copper	Page 103
Feldspar	Page 105
Fluorspar	Page 108
Gold	Page 110
Graphite.....	Page 112
Gypsum	Page 115
Helium	Page 117
Hydraulic Cement	Page 120
Iodine	Page 122
Iron	Page 125
Kyanite	Page 127
Lead	Page 130
Lithium	Page 132
Magnesite	Page 135
Magnesium.....	Page 137
Manganese	Page 140
Mercury	Page 142
Molybdenum	Page 144
Nickel	Page 147
Niobium	Page 149
Nitrogen	Page 152
Phosphate Rock	Page 154

Platinum	Page 157
Potash	Page 159
Rare Earths	Page 161
Salt	Page 164
Selenium	Page 166
Silicon	Page 168
Silver	Page 171
Sulfur	Page 173
Talc	Page 176
Tantalum	Page 178
Tin	Page 180
Titanium	Page 182
Tungsten.....	Page 185
Vanadium	Page 187
Zinc	Page 190
Zirconium	Page 192
Appendix 4 : Energy Sources Data	Page 195
U.S. Biofuel Energy Production	Page 195
U.S. Biomass Energy Production	Page 197
U.S. Coal Energy Production	Page 200
U.S. Fossil Fuel Energy Production	Page 202
U.S. Geothermal Energy Production	Page 205
U.S. Hydroelectric Energy Production	Page 207
U.S. Natural Gas Energy Production	Page 209
U.S. Nuclear Energy Production	Page 212

U.S. Oil Energy Production	Page 214
U.S. Renewable Energy Production	Page 216
U.S. Solar Energy Production	Page 219
U.S. Total Energy Production	Page 221
U.S. Wind Energy Production	Page 224
U.S. Wood Energy Production	Page 226
Appendix 5: Energy Materials Data	Page 229
U.S. Coal	Page 229
U.S. Natural Gas	Page 231
U.S. Oil	Page 234
U.S. Uranium Usage	Page 236
Appendix 6: Patent Search Keywords.....	Page 239
Appendix 7: Scaling	Page 241
Appendix 8: Executive Summary	Page 245
References	Page 260

List of Tables and Figures

Introduction

Table 1:	Selected Engineering Materials	Page 2
Table 2:	Selected Energy Sources	Page 2

Section 3: Patent and Production Activity Data Correlation

Figure 1 :	Rare Earths: Activity and Patents	Page 12
Figure 2:	Beryllium: Activity and Patents	Page 12
Table 3:	Overall Correlation Coefficients	Page 13

Section 4: Best-Fit

Figure 3 :	Typical Long-Term Life Cycle of a Metal	Page 15
Table 4 :	Common Pattern Equation Variables	Page 17
Figure 4 :	USGS World Chromium Production	Page 19
Figure 5 :	EPO Worldwide Patent Search: Chromium	Page 20
Table 5 :	Individual Origins, Origin Shifts, Stages, r and R^2	Page 21
Figure 6 :	Chromium: Best-Fit Activity and Patents	Page 23
Figure 7 :	Zinc: Best-Fit Activity and Patents	Page 23
Figure 8 :	Aluminum: Activity and Patents	Page 25
Figure 9 :	Arsenic: Activity and Patents	Page 26
Figure 10 :	Manganese: Activity and Patents	Page 26
Figure 11 :	Mercury: Activity and Patents	Page 27

Section 5: Best-Fit, Origin Shift and Innovation

Table 6 : α and n Parameters, α^n ratio, and Origin Shift Page 29

Figure 12(a) : Engineering Material Drive Ratio vs. Origin Shift Page 32

Figure 12(b) : Drive Ratio vs. Origin Shift (R^2 values above 0.85) Page 34

Table 7 : Origin Shifts and R^2 values Page 35

Figure 12(c) : Drive Ratio vs. Origin Shift (2 shift average) Page 36

Section 6: Energy Sources

Table 8 : Energy Source r and $100r^2$ Page 39

Figure 13 : U.S. Wind Energy: Activity and Patents Page 39

Figure 14 : U.S. Oil Energy: Activity and Patents Page 40

Figure 15 : Typical Energy Source Long-Term Life Cycle Page 40

Figure 16 : EIA U.S. Wind Energy Production Page 41

Figure 17 : EPO Worldwide Patent Search: Wind Power Page 42

Table 9 : Individual Origins, Origin Shifts, Stages, r' and R^2 Page 44

Figure 18 : U.S. Wind Energy Best-Fit Activity and Patents Page 45

Figure 19 : U.S. Geothermal Energy Best-Fit Activity and PatentsPage 45

Table 10: α and n parameters, α^n ratio and Origin Shifts Page 46

Figure 20 : Energy Source Drive Ratio vs. Origin Ratio Page 48

Table 11 : Energy Material R^2 values, correlation and origin shifts.....Page 50

Table 12 : Energy Material alpha and n parametersPage 50

Figure 21: Energy Material Drive Ratio vs. OriginPage 51

Table 13 : Energy Sources and Engineering Materials Page 52

Table 14 : Energy Source and Material ComparisonPage 52

Section 7: Analysis

Figure 22 : Mercury Production Life Cycle Page 57
Table 15 : Result ChangesPage 61
Figure 23 : Manganese Production Activity 1900-2004 Page 62
Figure 24 : Manganese Production Activity 1900-2007 Page 62

Appendix 1: Correlation

Table A1.1 : Calculated Correlation Terms Page 70
Figure A1.1 : Aluminum Activity and Patent Data Page 70

Appendix 2: Matlab

Program A2.1 : MatLab Best-Fit Template Page 71
Program A2.2 : Zinc Production MatLab Program Page 72
Program A2.4 : Zinc Program Results Page 74

Appendix 3: Engineering Materials Data

Table A3.1 : Aluminum Activity and Patents Page 76
Table A3.2 : Correlation Equation Terms Page 76
Figure A3.1 : Aluminum: Activity and Patents Page 76
Figure A3.2 : USGS World Aluminum Production Page 77
Figure A3.3 : EPO Worldwide Patent Search: Aluminum Page 77
Figure A3.4 : Aluminum Best-Fit Activity and Patents Page 77
Figure A3.5 : Aluminum Independent Patent Best-Fit Page 78
Table A3.3 : Antimony Activity and Patents Page 78
Table A3.4 : Correlation Equation Terms Page 79

Figure A3.6 :	Antimony: Activity and Patents	Page 79
Figure A3.7 :	USGS World Antimony Production	Page 79
Figure A3.8 :	EPO Worldwide Patent Search: Antimony	Page 80
Figure A3.9 :	Antimony Best-Fit Activity and Patents	Page 80
Figure A3.10 :	Antimony Independent Patent Best-Fit	Page 80
Table A3.5 :	Arsenic Activity and Patents	Page 81
Table A3.6 :	Correlation Equation Terms	Page 81
Figure A3.11 :	Arsenic: Activity and Patents	Page 81
Figure A3.12 :	USGS World Arsenic Production	Page 82
Figure A3.13:	Arsenic Independent Patent Best-Fit	Page 82
Table A3.7 :	Asbestos Activity and Patents	Page 83
Table A3.8 :	Correlation Equation Terms	Page 83
Figure A3.14 :	Asbestos: Activity and Patents	Page 83
Figure A3.15 :	USGS World Asbestos Production	Page 84
Figure A3.16:	Asbestos Independent Patent Best-Fit	Page 84
Table A3.9 :	Barite Activity and Patents	Page 85
Table A3.10 :	Correlation Equation Terms	Page 85
Figure A3.17 :	Barite: Activity and Patents	Page 85
Figure A3.18 :	USGS World Barite Production	Page 86
Figure A3.19 :	EPO Worldwide Patent Search: Barite	Page 86
Figure A3.20 :	Barite Best-Fit Activity and Patents	Page 86
Figure A3.21 :	Barite Independent Patent Best-Fit	Page 87
Table A3.11 :	Bauxite/Alumina Activity and Patents	Page 87
Table A3.12 :	Correlation Equation Terms	Page 88
Figure A3.22 :	Bauxite/Alumina: Activity and Patents	Page 88

Figure A3.23 :	USGS World Bauxite/Alumina Production	Page 88
Figure A3.24 :	EPO Worldwide Patent Search: Bauxite/Alumina	Page 89
Figure A3.25 :	Bauxite/Alumina Best-Fit Activity and Patents	Page 89
Figure A3.26:	Bauxite/Alumina Independent Patent Best-Fit	Page 89
Table A3.13 :	Beryllium Activity and Patents	Page 90
Table A3.14 :	Correlation Equation Terms	Page 90
Figure A3.27 :	Beryllium: Activity and Patents	Page 90
Figure A3.28 :	USGS World Beryllium Production	Page 91
Figure A3.29:	Beryllium Independent Patent Best-Fit	Page 91
Table A3.15 :	Bismuth Activity and Patents	Page 92
Table A3.16 :	Correlation Equation Terms	Page 92
Figure A3.30 :	Bismuth: Activity and Patents	Page 92
Figure A3.31 :	USGS World Bismuth Production	Page 93
Figure A3.32:	Bismuth Independent Patent Best-Fit	Page 93
Table A3.17 :	Boron Activity and Patents	Page 94
Table A3.18 :	Correlation Equation Terms	Page 94
Figure A3.33 :	Boron: Activity and Patents	Page 94
Figure A3.34 :	USGS World Boron Production	Page 95
Figure A3.35:	Boron Independent Patent Best-Fit	Page 95
Table A3.19 :	Cadmium Activity and Patents	Page 96
Table A3.20 :	Correlation Equation Terms	Page 96
Figure A3.36 :	Cadmium: Activity and Patents	Page 96
Figure A3.37 :	USGS World Cadmium Production	Page 97
Figure A3.38:	Cadmium Independent Patent Best-Fit	Page 97

Table A3.21 :	Chromium Activity and Patents	Page 98
Table A3.22 :	Correlation Equation Terms	Page 98
Figure A3.39 :	Chromium: Activity and Patents	Page 98
Figure A3.40 :	USGS World Chromium Production	Page 99
Figure A3.41 :	EPO Worldwide Patent Search: Chromium	Page 99
Figure A3.42 :	Chromium Best-Fit Activity and Patents	Page 99
Figure A3.43 :	Chromium Independent Patent Best-Fit	Page 100
Table A3.23 :	Cobalt Activity and Patents	Page 100
Table A3.24 :	Correlation Equation Terms	Page 101
Figure A3.44 :	Cobalt: Activity and Patents	Page 101
Figure A3.45 :	USGS World Cobalt Production	Page 101
Figure A3.46 :	EPO Worldwide Patent Search: Cobalt	Page 102
Figure A3.47 :	Cobalt Best-Fit Activity and Patents	Page 102
Figure A3.48 :	Cobalt Independent Patent Best-Fit	Page 102
Table A3.25 :	Copper Activity and Patents	Page 103
Table A3.26 :	Correlation Equation Terms	Page 103
Figure A3.49 :	Copper: Activity and Patents	Page 103
Figure A3.50 :	USGS World Copper Production	Page 104
Figure A3.51 :	EPO Worldwide Patent Search: Copper	Page 104
Figure A3.52 :	Copper Best-Fit Activity and Patents	Page 104
Figure A3.53 :	Copper Independent Patent Best-Fit	Page 105
Table A3.27 :	Feldspar Activity and Patents	Page 105
Table A3.28 :	Correlation Equation Terms	Page 106
Figure A3.54 :	Feldspar: Activity and Patents	Page 106
Figure A3.55 :	USGS World Feldspar Production	Page 106

Figure A3.56 :	EPO Worldwide Patent Search: Feldspar	Page 107
Figure A3.57 :	Feldspar Best-Fit Activity and Patents	Page 107
Figure A3.58:	Feldspar Independent Patent Best-Fit	Page 107
Table A3.29 :	Fluorspar Activity and Patents	Page 108
Table A3.30 :	Correlation Equation Terms	Page 108
Figure A3.59 :	Fluorspar: Activity and Patents	Page 108
Figure A3.60 :	USGS World Fluorspar Production	Page 109
Figure A3.61 :	EPO Worldwide Patent Search: Fluorspar	Page 109
Figure A3.62 :	Fluorspar Best-Fit Activity and Patents	Page 109
Figure A3.63:	Fluorspar Independent Patent Best-Fit	Page 110
Table A3.31 :	Gold Activity and Patents	Page 110
Table A3.32 :	Correlation Equation Terms	Page 111
Figure A3.64 :	Gold: Activity and Patents	Page 111
Figure A3.65 :	USGS World Gold Production	Page 111
Figure A3.66:	Gold Independent Patent Best-Fit	Page 112
Table A3.33 :	Graphite Activity and Patents	Page 112
Table A3.34 :	Correlation Equation Terms	Page 113
Figure A3.67 :	Graphite: Activity and Patents	Page 113
Figure A3.68 :	USGS World Graphite Production	Page 113
Figure A3.69 :	EPO Worldwide Patent Search: Graphite	Page 114
Figure A3.70 :	Graphite Best-Fit Activity and Patents	Page 114
Figure A3.71 :	Graphite Independent Patent Best-Fit	Page 114
Table A3.35 :	Gypsum Activity and Patents	Page 115
Table A3.36 :	Correlation Equation Terms	Page 115

Figure A3.72 :	Gypsum: Activity and Patents	Page 115
Figure A3.73 :	USGS World Gypsum Production	Page 116
Figure A3.74 :	EPO Worldwide Patent Search: Gypsum	Page 116
Figure A3.75 :	Gypsum Best-Fit Activity and Patents	Page 116
Figure A3.76:	Gypsum Independent Patent Best-Fit	Page 117
Table A3.37 :	Helium Activity and Patents	Page 117
Table A3.38 :	Correlation Equation Terms	Page 118
Figure A3.77 :	Helium: Activity and Patents	Page 118
Figure A3.78 :	USGS World Helium Production	Page 118
Figure A3.79 :	EPO Worldwide Patent Search: Helium	Page 119
Figure A3.80 :	Helium Best-Fit Activity and Patents	Page 119
Figure A3.81 :	Helium Independent Patent Best-Fit	Page 119
Table A3.39 :	Hydraulic Cement Activity and Patents	Page 120
Table A3.40 :	Correlation Equation Terms	Page 120
Figure A3.82 :	Hydraulic Cement: Activity and Patents	Page 120
Figure A3.83 :	USGS World Hydraulic Cement Production	Page 121
Figure A3.84 :	EPO Worldwide Patent Search: Hydraulic Cement	Page 121
Figure A3.85 :	Hydraulic Cement Best-Fit Activity and Patents	Page 121
Figure A3.86:	Hydraulic Cement Independent Patent Best-Fit	Page 122
Table A3.41 :	Iodine Activity and Patents	Page 122
Table A3.42 :	Correlation Equation Terms	Page 123
Figure A3.87 :	Iodine: Activity and Patents	Page 123
Figure A3.88 :	USGS World Iodine Production	Page 123
Figure A3.89 :	EPO Worldwide Patent Search: Iodine	Page 124
Figure A3.90 :	Iodine Best-Fit Activity and Patents	Page 124

Figure A3.91 :	Iodine Independent Patent Best-Fit	Page 124
Table A3.43 :	Iron Activity and Patents	Page 125
Table A3.44 :	Correlation Equation Terms	Page 125
Figure A3.92 :	Iron: Activity and Patents	Page 125
Figure A3.93 :	USGS World Iron Production	Page 126
Figure A3.94 :	EPO Worldwide Patent Search: Iron	Page 126
Figure A3.95 :	Iron Best-Fit Activity and Patents	Page 126
Figure A3.96:	Iron Independent Patent Best-Fit	Page 127
Table A3.45 :	Kyanite Activity and Patents	Page 127
Table A3.46 :	Correlation Equation Terms	Page 128
Figure A3.97 :	Kyanite: Activity and Patents	Page 128
Figure A3.98 :	USGS World Kyanite Production	Page 128
Figure A3.99 :	EPO Worldwide Patent Search: Kyanite	Page 129
Figure A3.100 :	Kyanite Best-Fit Activity and Patents	Page 129
Figure A3.101 :	Kyanite Independent Patent Best-Fit	Page 129
Table A3.47 :	Lead Activity and Patents	Page 130
Table A3.48 :	Correlation Equation Terms	Page 130
Figure A3.102 :	Lead: Activity and Patents	Page 130
Figure A3.103 :	USGS World Lead Production	Page 131
Figure A3.104 :	EPO Worldwide Patent Search: Lead	Page 131
Figure A3.105 :	Lead Best-Fit Activity and Patents	Page 131
Figure A3.106 :	Lead Independent Patent Best-Fit	Page 132
Table A3.49 :	Lithium Activity and Patents	Page 132
Table A3.50 :	Correlation Equation Terms	Page 133

Figure A3.107 :	Lithium: Activity and Patents	Page 133
Figure A3.108 :	USGS World Lithium Production	Page 133
Figure A3.109 :	EPO Worldwide Patent Search: Lithium	Page 134
Figure A3.110 :	Lithium Best-Fit Activity and Patents	Page 134
Figure A3.111 :	Lithium Independent Patent Best-Fit	Page 134
Table A3.51 :	Magnesite Activity and Patents	Page 135
Table A3.52 :	Correlation Equation Terms	Page 135
Figure A3.112 :	Magnesite: Activity and Patents	Page 135
Figure A3.113 :	USGS World Magnesite Production	Page 136
Figure A3.114 :	EPO Worldwide Patent Search: Magnesite	Page 136
Figure A3.115 :	Magnesite Best-Fit Activity and Patents	Page 136
Figure A3.116 :	Magnesite Independent Patent Best-Fit	Page 137
Table A3.53 :	Magnesium Activity and Patents	Page 137
Table A3.54 :	Correlation Equation Terms	Page 138
Figure A3.117 :	Magnesium: Activity and Patents	Page 138
Figure A3.118 :	USGS World Magnesium Production	Page 138
Figure A3.119 :	EPO Worldwide Patent Search: Magnesium	Page 139
Figure A3.120 :	Magnesium Best-Fit Activity and Patents	Page 139
Figure A3.121 :	Magnesium Independent Patent Best-Fit	Page 139
Table A3.55 :	Manganese Activity and Patents	Page 140
Table A3.56 :	Correlation Equation Terms	Page 140
Figure A3.122 :	Manganese: Activity and Patents	Page 140
Figure A3.123 :	USGS World Manganese Production	Page 141
Figure A3.124 :	EPO Worldwide Patent Search: Manganese	Page 141
Figure A3.125 :	Manganese Best-Fit Activity and Patents	Page 141

Figure A3.126 :	Manganese Independent Patent Best-Fit	Page 142
Table A3.57 :	Mercury Activity and Patents	Page 142
Table A3.58 :	Correlation Equation Terms	Page 143
Figure A3.127 :	Mercury: Activity and Patents	Page 143
Figure A3.128 :	USGS World Mercury Production	Page 143
Figure A3.129:	Mercury Independent Patent Best-Fit	Page 144
Table A3.59 :	Molybdenum Activity and Patents	Page 144
Table A3.60:	Correlation Equation Terms	Page 145
Figure A3.130 :	Molybdenum: Activity and Patents	Page 145
Figure A3.131 :	USGS World Molybdenum Production	Page 145
Figure A3.132 :	EPO Worldwide Patent Search: Molybdenum	Page 146
Figure A3.133 :	Molybdenum Best-Fit Activity and Patents	Page 146
Figure A3.134:	Molybdenum Independent Patent Best-Fit	Page 146
Table A3.61 :	Nickel Activity and Patents	Page 147
Table A3.62 :	Correlation Equation Terms	Page 147
Figure A3.135 :	Nickel: Activity and Patents	Page 147
Figure A3.136 :	USGS World Nickel Production	Page 148
Figure A3.137 :	EPO Worldwide Patent Search: Nickel	Page 148
Figure A3.138 :	Nickel Best-Fit Activity and Patents	Page 148
Figure A3.139 :	Nickel Independent Patent Best-Fit	Page 149
Table A3.63 :	Niobium Activity and Patents	Page 149
Table A3.64 :	Correlation Equation Terms	Page 150
Figure A3.140 :	Niobium: Activity and Patents	Page 150
Figure A3.141 :	USGS World Niobium Production	Page 150

Figure A3.142 :	EPO Worldwide Patent Search: Niobium	Page 151
Figure A3.143 :	Niobium Best-Fit Activity and Patents	Page 151
Figure A3.144 :	Niobium Independent Patent Best-Fit	Page 151
Table A3.65 :	Nitrogen Activity and Patents	Page 152
Table A3.66 :	Correlation Equation Terms	Page 152
Figure A3.145 :	Nitrogen: Activity and Patents	Page 152
Figure A3.146 :	USGS World Nitrogen Production	Page 153
Figure A3.147 :	EPO Worldwide Patent Search: Nitrogen	Page 153
Figure A3.148 :	Nitrogen Best-Fit Activity and Patents	Page 153
Figure A3.149:	Nitrogen Independent Patent Best-Fit	Page 154
Table A3.67 :	Phosphate Rock Activity and Patents	Page 154
Table A3.68 :	Correlation Equation Terms	Page 155
Figure A3.150 :	Phosphate Rock: Activity and Patent	Page 155
Figure A3.151 :	USGS World Phosphate Rock Production	Page 155
Figure A3.152 :	EPO Worldwide Patent Search: Phosphate Rock	Page 156
Figure A3.153 :	Phosphate Rock Best-Fit Activity and Patents	Page 156
Figure A3.154:	Phosphate Rock Independent Patent Best-Fit	Page 156
Table A3.69 :	Platinum Activity and Patents	Page 157
Table A3.70 :	Correlation Equation Terms	Page 157
Figure A3.155 :	Platinum: Activity and Patents	Page 157
Figure A3.156 :	USGS World Platinum Production	Page 158
Figure A3.157 :	EPO Worldwide Patent Search: Platinum	Page 158
Figure A3.158 :	Platinum Best-Fit Activity and Patents	Page 158
Figure A3.159:	Platinum Independent Patent Best-Fit	Page 159
Table A3.71 :	Potash Activity and Patents	Page 159

Table A3.72 :	Correlation Equation Terms	Page 160
Figure A3.160 :	Potash: Activity and Patents	Page 160
Figure A3.161:	USGS World Potash Production	Page 160
Figure A3.162:	Potash Independent Patent Best-Fit	Page 161
Table A3.73 :	Rare earths Activity and Patents	Page 161
Table A3.74 :	Correlation Equation Terms	Page 162
Figure A3.163 :	Rare Earths: Activity and Patents	Page 162
Figure A3.164 :	USGS World Rare Earths Production.....	Page 162
Figure A3.165 :	EPO Worldwide Patent Search: Rare Earths	Page 163
Figure A3.166 :	Rare Earths Best-Fit Activity and Patents	Page 163
Figure A3.167 :	Rare Earths Independent Patent Best-Fit	Page 163
Table A3.75 :	Salt Activity and Patents	Page 164
Table A3.76 :	Correlation Equation Terms	Page 164
Figure A3.168 :	Salt: Activity and Patents	Page 164
Figure A3.169:	USGS World Salt Production	Page 165
Figure A3.170 :	EPO Worldwide Patent Search: Salt	Page 165
Figure A3.171 :	Salt Best-Fit Activity and Patents	Page 165
Figure A3.172:	Salt Independent Patent Best-Fit	Page 166
Table A3.77 :	Selenium Activity and Patents	Page 166
Table A3.78 :	Correlation Equation Terms	Page 167
Figure A3.173 :	Selenium: Activity and Patents	Page 167
Figure A3.174:	USGS World Selenium Production	Page 167
Figure A3.175:	Selenium Independent Patent Best-Fit	Page 168
Table A3.79 :	Silicon Activity and Patents	Page 168

Table A3.80 :	Correlation Equation Terms	Page 169
Figure A3.176 :	Silicon: Activity and Patents	Page 169
Figure A3.177 :	USGS World Silicon Production	Page 169
Figure A3.178 :	EPO Worldwide Patent Search: Silicon	Page 170
Figure A3.179 :	Silicon Best-Fit Activity and Patents	Page 170
Figure A3.180 :	Silicon Independent Patent Best-Fit	Page 170
Table A3.81 :	Silver Activity and Patents	Page 171
Table A3.82 :	Correlation Equation Terms	Page 171
Figure A3.181 :	Silver: Activity and Patents	Page 171
Figure A3.182:	USGS World Silver Production	Page 172
Figure A3.183 :	EPO Worldwide Patent Search: Silver	Page 172
Figure A3.184:	Silver Best-Fit Activity and Patents	Page 172
Figure A3.185:	Silver Independent Patent Best-Fit	Page 173
Table A3.83 :	Sulfur Activity and Patents	Page 173
Table A3.84 :	Correlation Equation Terms	Page 174
Figure A3.186 :	Sulfur: Activity and Patents	Page 174
Figure A3.187:	USGS World Sulfur Production	Page 174
Figure A3.188 :	EPO Worldwide Patent Search: Sulfur	Page 175
Figure A3.189 :	Sulfur Best-Fit Activity and Patents	Page 175
Figure A3.190:	Sulfur Independent Patent Best-Fit	Page 175
Table A3.85 :	Talc Activity and Patents	Page 176
Table A3.86 :	Correlation Equation Terms	Page 176
Figure A3.191 :	Talc: Activity and Patents	Page 176
Figure A3.192:	USGS World Talc Production	Page 177
Figure A3.193 :	EPO Worldwide Patent Search: Talc	Page 177

Figure A3.194 :	Talc Best-Fit Activity and Patents	Page 177
Figure A3.195:	Talc Independent Patent Best-Fit	Page 178
Table A3.87 :	Tantalum Activity and Patents	Page 178
Table A3.88 :	Correlation Equation Terms	Page 179
Figure A3.196 :	Tantalum: Activity and Patents	Page 179
Figure A3.197 :	USGS World Tantalum Production	Page 179
Figure A3.198:	Tantalum Independent Patent Best-Fit	Page 180
Table A3.89 :	Tin Activity and Patents	Page 180
Table A3.90 :	Correlation Equation Terms	Page 181
Figure A3.199 :	Tin: Activity and Patents	Page 181
Figure A3.200:	USGS World Tin Production	Page 181
Figure A3.201:	Tin Independent Patent Best-Fit	Page 182
Table A3.91 :	Titanium Activity and Patents	Page 182
Table A3.92 :	Correlation Equation Terms	Page 183
Figure A3.202 :	Titanium: Activity and Patents	Page 183
Figure A3.203 :	USGS World Titanium Production	Page 183
Figure A3.204 :	EPO Worldwide Patent Search: Titanium	Page 184
Figure A3.205 :	Titanium Best-Fit Activity and Patents	Page 184
Figure A3.206 :	Titanium Independent Patent Best-Fit	Page 184
Table A3.93 :	Tungsten Activity and Patents	Page 185
Table A3.94 :	Correlation Equation Terms	Page 185
Figure A3.207 :	Tungsten: Activity and Patents	Page 185
Figure A3.208 :	USGS World Tungsten Production	Page 186
Figure A3.209 :	EPO Worldwide Patent Search: Tungsten	Page 186

Figure A3.210:	Tungsten Best-Fit Activity and Patents	Page 186
Figure A3.211 :	Tungsten Independent Patent Best-Fit	Page 187
Table A3.95 :	Vanadium Activity and Patents	Page 187
Table A3.96 :	Correlation Equation Terms	Page 188
Figure A3.212 :	Vanadium: Activity and Patents	Page 188
Figure A3.213 :	USGS World Vanadium Production	Page 188
Figure A3.214 :	EPO Worldwide Patent Search: Vanadium	Page 189
Figure A3.215 :	Vanadium Best-Fit Activity and Patents	Page 189
Figure A3.216:	Vanadium Independent Patent Best-Fit	Page 189
Table A3.97 :	Zinc Activity and Patents	Page 190
Table A3.98 :	Correlation Equation Terms	Page 190
Figure A3.217 :	Zinc: Activity and Patents	Page 190
Figure A3.218 :	USGS World Zinc Production	Page 191
Figure A3.219 :	EPO Worldwide Patent Search: Zinc	Page 191
Figure A3.220 :	Zinc Best-Fit Activity and Patents	Page 191
Figure A3.221 :	Zinc Independent Patent Best-Fit	Page 192
Table A3.99 :	Zirconium Activity and Patents	Page 192
Table A3.100 :	Correlation Equation Terms	Page 193
Figure A3.222 :	Zirconium: Activity and Patents	Page 193
Figure A3.223 :	USGS World Zirconium Production	Page 193
Figure A3.224 :	EPO Worldwide Patent Search: Zirconium	Page 194
Figure A3.225 :	Zirconium Best-Fit Activity and Patents	Page 194
Figure A3.226 :	Zirconium Independent Patent Best-Fit	Page 194

Appendix 4: Energy Sources Data

Table A4.1 :	U.S. Biofuel Energy Activity and Patents	Page 195
Table A4.2 :	Correlation Equation Terms	Page 195
Figure A4.1 :	U.S. Biofuel energy: Activity and Patents	Page 195
Figure A4.2 :	EIA U.S. Biofuel Energy Production	Page 196
Figure A4.3 :	EPO Worldwide Patent Search: Biofuel Energy	Page 196
Figure A4.4 :	U.S. Biofuel Energy Best-Fit Activity and Patents	Page 196
Figure A4.5 :	U.S. Biofuel Energy Independent Patent Best-Fit	Page 197
Table A4.3 :	U.S. Biomass Energy Activity and Patents	Page 197
Table A4.4 :	Correlation Equation Terms	Page 198
Figure A4.6 :	U.S. Biomass Energy: Activity and Patents	Page 198
Figure A4.7 :	EIA U.S. Biomass Energy Production	Page 198
Figure A4.8 :	EPO Worldwide Patent Search: Energy Power	Page 199
Figure A4.9 :	U.S. Biomass Energy Best-Fit Activity and Patents	Page 199
Figure A4.10 :	U.S. Biomass Energy Independent Patent Best-Fit	Page 199
Table A4.5 :	U.S. Coal Energy Activity and Patents	Page 200
Table A4.6 :	Correlation Equation Terms	Page 200
Figure A4.11 :	U.S. Coal Energy: Activity and Patents	Page 200
Figure A4.12 :	EIA U.S. Coal Energy Production	Page 201
Figure A4.13 :	EPO Worldwide Patent Search: Coal Energy Production	Page 201
Figure A4.14 :	Coal Energy Production Best-Fit Activity and Patents	Page 201
Figure A4.15 :	U.S. Coal Energy Independent Patent Best-Fit	Page 202
Table A4.7 :	U.S. Fossil Fuel Energy Activity and Patents	Page 202
Table A4.8 :	Correlation Equation Terms	Page 203
Figure A4.16 :	U.S. Fossil Fuel Energy: Activity and Patents	Page 203

Figure A4.17 :	EIA U.S. Fossil Fuel Energy Production Page 203
Figure A4.18 :	EPO Worldwide Patent Search: Fossil Fuel Energy Page 204
Figure A4.19 :	U.S. Fossil Fuel Energy Best-Fit Activity and PatentsPage 204
Figure A4.20 :	U.S. Fossil Fuel Energy Independent Patent Best-Fit.....	Page 204
Table A4.9 :	U.S. Geothermal Energy Activity and Patents Page 205
Table A4.10 :	Correlation Equation Terms Page 205
Figure A4.21 :	U.S. Geothermal Energy: Activity and Patents Page 205
Figure A4.22 :	EIA U.S. Geothermal Energy Production Page 206
Figure A4.23 :	EPO Worldwide Patent Search: Geothermal EnergyPage 206
Figure A4.24 :	U.S. Geothermal Energy Best-Fit Activity and Patents	...Page 206
Figure A4.25:	U.S. Geothermal Energy Independent Patent Best-Fit Page 207
Table A4.11 :	U.S. Hydroelectric Energy Activity and Patents Page 207
Table A4.12 :	Correlation Equation Terms Page 208
Figure A4.26 :	U.S. Hydroelectric Energy: Activity and Patents Page 208
Figure A4.27 :	EIA U.S. Hydroelectric Energy Production Page 208
Figure A4.28 :	U.S. Hydroelectric Energy Independent Patent Best-Fit Page 209
Table A4.13 :	U.S. Natural Gas Energy Activity and Patents Page 209
Table A4.14 :	Correlation Equation Terms Page 210
Figure A4.29 :	U.S. Natural Gas Energy: Activity and Patents Page 210
Figure A4.30 :	EIA U.S. Natural Gas Energy Production Page 210
Figure A4.31 :	EPO Worldwide Patent Search: Natural Gas Energy Page 211
Figure A4.32 :	Natural Gas Energy Best-Fit Activity and Patents Page 211
Figure A4.33 :	U.S. Natural Gas Energy Independent Patent Best-Fit.....	Page 211
Table A4.15 :	U.S. Nuclear Energy Activity and Patents Page 212
Table A4.16 :	Correlation Equation Terms Page 212

Figure A4.34 :	U.S. Nuclear Energy: Activity and Patents	Page 212
Figure A4.35 :	EIA U.S. Nuclear Energy Production	Page 213
Figure A4.36 :	EPO Worldwide Patent Search: Nuclear Energy	Page 213
Figure A4.37 :	U.S. Nuclear Energy Best-Fit Activity and Patents	Page 213
Figure A4.38 :	U.S. Nuclear Energy Independent Patent Best-Fit	Page 214
Table A4.17 :	U.S. Oil Energy Activity and Patents	Page 214
Table A4.18 :	Correlation Equation Terms	Page 215
Figure A4.39 :	U.S. Oil Energy: Activity and Patents	Page 215
Figure A4.40 :	EIA U.S. Oil Energy Production	Page 215
Figure A4.41 :	U.S. Oil Energy Independent Patent Best-Fit	Page 216
Table A4.19 :	U.S. Renewable Energy Activity and Patents	Page 216
Table A4.20 :	Correlation Equation Terms	Page 217
Figure A4.42 :	U.S. Renewable Energy: Activity and Patents	Page 217
Figure A4.43 :	EIA U.S. Renewable Energy Production	Page 217
Figure A4.44 :	EPO Worldwide Patent Search: Renewable Energy	Page 218
Figure A4.45 :	U.S. Renewable Energy Best-Fit Activity and Patents	Page 218
Figure A4.46 :	U.S. Renewable Energy Independent Patent Best-Fit	Page 218
Table A4.21 :	U.S. Solar Energy Activity and Patents	Page 219
Table A4.22 :	Correlation Equation Terms	Page 219
Figure A4.47 :	U.S. Solar Energy: Activity and Patents	Page 219
Figure A4.48 :	EIA U.S. Solar Energy Production	Page 220
Figure A4.49 :	EPO Worldwide Patent Search: Solar Energy	Page 220
Figure A4.50 :	U.S. Solar Energy Best-Fit Activity and Patents	Page 220
Figure A4.51 :	U.S. Solar Energy Independent Patent Best-Fit	Page 221

Table A4.23 :	U.S. Total Energy Activity and Patents	Page 221
Table A4.24 :	Correlation Equation Terms	Page 222
Figure A4.52 :	U.S. Total Energy: Activity and Patents	Page 222
Figure A4.53:	EIA U.S. Total Energy Production	Page 222
Figure A4.54 :	EPO Worldwide Patent Search: Total Energy	Page 223
Figure A4.55 :	U.S. Total Energy Best-Fit Activity and Patents	Page 223
Figure A4.56:	U.S. Total Energy Independent Patent Best-Fit	Page 223
Table A4.27 :	U.S. Wind Power Activity and Patents	Page 224
Table A4.28 :	Correlation Equation Terms	Page 224
Figure A4.57 :	U.S. Wind Energy: Activity and Patents	Page 224
Figure A4.58 :	EIA U.S. Wind Energy Production	Page 225
Figure A4.59 :	EPO Worldwide Patent Search: Wind Energy	Page 225
Figure A4.60 :	U.S. Wind Energy Best-Fit Activity and Patents	Page 225
Figure A4.61:	U.S. Wind Energy Independent Patent Best-Fit	Page 226
Table A4.29 :	U.S. Wood Energy Activity and Patents	Page 226
Table A4.30 :	Correlation Equation Terms	Page 227
Figure A4.66 :	U.S. Wood Energy: Activity and Patents	Page 227
Figure A4.67:	EIA U.S. Wood Energy Production	Page 227
Figure A4.68:	U.S. Wood Energy Independent Patent Best-Fit	Page 228

Appendix 5: EnergyMaterials Data

Table A5.1 :	U.S. Coal Activity and Patents	Page 229
Table A5.2 :	Correlation Equation Terms	Page 229
Figure A5.1 :	U.S. Coal : Activity and Patents	Page 229

Figure A5.2 :	EIA U.S. Coal Production	Page 230
Figure A5.3 :	EPO Worldwide Patent Search: Coal Production	Page 230
Figure A5.4 :	Coal Production Best-Fit Activity and Patents	Page 230
Figure A5.5 :	U.S. Coal Independent Patent Best-Fit	Page 231
Table A5.3 :	U.S. Natural Gas Activity and Patents	Page 231
Table A5.4 :	Correlation Equation Terms	Page 232
Figure A5.6 :	U.S. Natural Gas : Activity and Patents	Page 232
Figure A5.7 :	EIA U.S. Natural Gas Production	Page 232
Figure A5.8 :	EPO Worldwide Patent Search: natural Gas Production	...	Page 233
Figure A5.9 :	Natural Gas Production Best-Fit Activity and Patents	Page 233
Figure A5.10 :	U.S. Natural Gas Independent Patent Best-Fit	Page 233
Table A5.5 :	U.S. Oil Activity and Patents	Page 234
Table A5.6 :	Correlation Equation Terms	Page 234
Figure A5.11 :	U.S. Oil : Activity and Patents	Page 234
Figure A5.12 :	EIA U.S. Oil Production	Page 235
Figure A5.13 :	EPO Worldwide Patent Search: Oil Production	Page 235
Figure A5.14 :	Oil Production Best-Fit Activity and Patents	Page 235
Figure A5.15 :	U.S. Oil Independent Patent Best-Fit	Page 236
Table A5.7 :	U.S. Uranium Activity and Patents	Page 236
Table A5.8 :	Correlation Equation Terms	Page 237
Figure A5.16 :	U.S. Uranium : Activity and Patents	Page 237
Figure A5.17 :	EIA U.S. Uranium Usage	Page 237
Figure A5.18 :	EPO Worldwide Patent Search: Uranium Usage	Page 238
Figure A5.19 :	Uranium Usage Best-Fit Activity and Patents	Page 238
Figure A5.20 :	U.S. Uranium Independent Patent Best-Fit	Page 238

Appendix 6: Patent Search Keywords

Table A6.1: Patent Search KeywordsPage 239

Appendix 7: Scaling

Table A7.1: Scale Materials and Sources ComparisonPage 242

Figure A7.1 : Bauxite Unscaled Production Best-FitPage 243

Figure A7.2 : Bauxite Scaled Production Best-FitPage 244

Appendix 8: Executive Summary

Figure A8.1: Illustration of a Typical Long-term Life Cycle for a Material.....Page 246

Figure A8.2: USGS World Chromium Production Page 246

Figure A8.3: Biofuel Energy Sensitivity Curve Page 250

Figure A8.4: Zinc Best-Fit Activity and PatentsPage 251

Figure A8.5: Engineering Materials Origin vs. Drive RatioPage 253

Figure A8.6: Energy Source Origin Ratio vs. Drive OriginPage 256

Table A8.1: Energy Sources and Related MaterialsPage 257

Table A8.2: Origin Shift, Origin Ratio and Drive RatioPage 257

Introduction

This dissertation continues a valuable and timely study concerned with the relationship between the production and innovation of materials. Invention is the realization and development of new and original ideas and products, while innovation is the successful utilization of such ideas and products, as well as means to conduct business, to market, and to finance, with the ultimate goal of making a profit. Innovations are often much more than inventions. Invention is the creative act or flash of genius while innovation is the exploitation of, and change caused by, the invention itself. Most inventions are technical, but innovations do not have to be technical at all, since, for example, technology is not necessary for the development of market or business model innovations [1,2]. This work examines the linkage between patent and production life cycles for various engineering materials (metals and non-metals) and explores whether such relationships also apply to energy sources and how such relationships can be employed as a possible predictive tool for the more efficient use and development of these materials and sources for, ultimately, a greater profit.

Invention is necessary for innovation to occur, but invention by itself is not enough for innovation to take place. Innovation can be described as being multi-dimensional, in that innovation requires vision concerning the invention, market need, timing, technology convergence and an implementation strategy [2]. Inventions are relatively low-risk with technology and intellectual property issues dominating. Innovations have large risks attached to them and are dominated by marketplace effectiveness, cost and profit concerns [2]. Anyone with a good idea and imagination can invent, but it takes someone with foresight, knowledge and courage to innovate effectively.

This dissertation expands the previously published work [1-3] on life cycle best-fit analysis to over fifty engineering materials, as well as fourteen energy sources. The selected engineering materials are listed in Table 1 and were not only chosen for the availability of

complete sets of production and patent data between the years of 1900-2007, but also as representatives of a wide variety of materials and their applications. Likewise, Table 2 presents fourteen energy sources that were chosen for their representative value as well as the availability of complete production and data sets for the years 1900-2008.

Table 1. Engineering Materials chosen for this study.

Aluminum	Chromium	Iodine	Nickel	Silver
Antimony	Cobalt	Iron	Niobium	Sulfur
Arsenic	Copper	Kyanite	Nitrogen	Talc
Asbestos	Feldspar	Lead	Phosphate	Tantalum
Barite	Fluorspar	Lithium	Platinum	Tin
Bauxite/Alumina	Gold	Magnesite	Potash	Titanium
Beryllium	Graphite	Magnesium	Rare Earths	Tungsten
Bismuth	Gypsum	Manganese	Salt	Vanadium
Boron	Helium	Mercury	Selenium	Zinc
Cadmium	Hydraulic Cement	Molybdenum	Silicon	Zirconium

Table 2. Energy sources chosen for this study.

U.S. Biofuel Energy	U.S. Hydroelectric Energy	U.S. Solar Energy
U.S. Biomass Energy	U.S. Natural Gas Energy	U.S. Total Energy
U.S. Coal Energy	U.S. Nuclear Energy	U.S. Wind Energy
U.S. Fossil Fuel Energy	U.S. Oil Energy	U.S. Wood Energy
U.S. Geothermal Energy	U.S. Renewable Energy	

A major goal of this work is to discover relationships between patents (technological inventions) and how such relationships affect, or are affected by, material production. During the 21st Century, which is a period of knowledge driven economies, it has been demonstrated that intellectual property could be a dominant force providing the capital that will continue to drive future worldwide economic growth [1,4-54]. However, there is no clear quantitative connection that has been established between patents and production excepted for limited work [1-3]. In order to perform such investigations, innovation needs to be defined and then a measurable proxy for it must be found. Sekhar et. al. [2,3] have clearly demarcated the

difference between invention and innovation based on the overall long-term life cycle. Here, further correlations are made between the production innovation activity, i.e., Stage III and patent activity in a similar Stage III or beyond for the overall life cycle for both measures. Several other methods have been employed to measure and define innovation starting with the work of Joseph Schumpeter [55-57].

Correlation theory was originally applied to the above listed engineering materials to determine if a relationship exists between the production data and the patent data. Best-fit analysis was then applied to the production data sets to generate the life cycles of each material, and then to the patent data sets to discover if any origin shifts in the equation result. Such an origin shift would indicate a driving force being present for the innovative activity that could be ascribed to the patent activity [1-3]. Here is sought a dividing line, possibly numerical, between the creative and destructive modes of the innovative process, which define the driving and driven behavior of innovation. Best-fit analysis will also be applied to determine if a method can be developed to positively identify Stage III materials by seeking common trends in patent and production data and resulting origin shifts in the life cycles of the data.

The procedures applied to engineering materials will then be applied to energy sources to determine if such resources have similar four-stage life cycles, driving force behavior and origin shifts produced by best-fit analysis, and in general follow similar patterns as do engineering materials. Similar behavior will permit already proven engineering material analyses to be applied to energy sources and may reveal relationships between energy sources and related materials that will allow predictions concerning future innovative growth and can lead to a more efficient and profitable allocation of ever more scarce natural and monetary resources.

Section 1: Measuring Innovation

Innovation. An important figure in modern attitudes toward innovation and its measurement is the economist Joseph Alois Schumpeter (1883-1950) [55]. Schumpeterian theory perceives the importance of innovation, and suggests that it is a central part of capitalist economies. This theory postulates that innovation propels the economy, which is in a state of constant change [55]. Capitalism is defined by an ebb and flow with cycles existing in it, which need to be evaluated using the historical record [55]. Innovation destroys and causes havoc as it builds anew. Old conditions and ways of thinking and acting are destroyed when innovations introduce new ideas, making the innovative act a double-edged sword. Entrepreneurs, called “New Men” drive innovation by making creative responses to change, in the form of innovative acts [55].

Schumpeter’s writings define his concept of innovation or “new combinations” as carried out by the entrepreneur or “new man” in the following manner:

This concept covers the following five cases: (1) The introduction of a new good – that is one which consumers are not yet familiar – or of a new quality of good. (2) The introduction of a new method of production, that is one not yet tested by experience in the branch of manufacture concerned, which need by no means be founded upon a discovery scientifically new, and can also exist in a new way of handling a commodity commercially. (3) The opening of a new market, that is a market into which the particular branch of manufacture of the country in question has not previously entered, whether or not this market has existed before. (4) The conquest of a new source of supply of raw materials or half-manufactured goods, again irrespective of whether this source already exists or whether it has first to be created. (5) The carrying out of the new organization of any industry, like the creation of a monopoly position (for example through trustification) or the breaking up of a monopoly position [56].

Or put more simply, an innovation is an invention that becomes economically successful and earns profit, where the invention is the creation and establishment of something new [57].

The prime motivation for the innovator in implementing the above is entrepreneurial profit. “When other participants in the same industry see the new level of high profit, they

quickly try to imitate the innovation. The entrepreneur tries to preserve his high profit for as long as possible, through patents, further innovation, secret processes and advertising – each move an act of ‘aggression directed against actual and would-be competitors [55].’” The process of incessant revolution of the economic structure from within by destruction of the old system and creation of a new one is titled “Creative Destruction [56].” Innovation is used to make profits, and in doing so great change occurs where new ways are created and old are cast aside. Schumpeterian theory stresses the centrality and importance of innovation in the economy, but methods are needed to quantify innovation to make it a useful indicator of present and future economic growth.

Measurement of Innovation. There is truly a multi-disciplinary interest in innovation and its quantification, with multiple books and articles being written on the subject in many diverse areas of research [58-77]. Measurement of innovation has proven to be a difficult task with much argument and difference of opinion, especially when patents are used as an indicator of innovation. Innovation measurement using patent data, and alternatively, innovation measurement excluding patent data, will be discussed below.

Measurements Excluding Patents. The belief has been put forward that patents may not be a reliable or representative measurement of innovation and that patents may even hamper innovation itself [58-72]. Some suggest that patents of dubious quality end up at the heart of legal disputes making everyone pay more for innovation and making it less likely to occur [58]. Fewer products in the marketplace are the result, since companies decide not to innovate with new products [58]. Other reports suggest that patents inhibit the innovative process by restricting other people’s creativity or that the costs of patenting could be used better elsewhere by businesses and that these costs are burdens on emerging businesses [59]. Patent

infringement suits are considered by some as evidence of a patent system gone wild leading to the stifling of innovation [60]. Some studies have also indicated that stronger, or broader patents do not increase innovation [61,62].

Even those who support the use of patent data have pointed out several direct problems with the use of patents as indicators of innovation. The reliability of patent measurement of innovation has been questioned since not all patented inventions prove to be innovations, many innovations are never patented and patents differ greatly in their economic impact [63]. On account of the differences in national patent offices, the interest in patenting by inventors differs between countries, and firms more often patent domestically rather than in foreign patent offices [64]. Patents do not always represent commercially exploited innovation and seem to be better used as representative of an input into the innovation rather than an output evidence of it [65].

There are no shortages in the literature of alternative innovation measurement techniques. One suggested method of innovation measurement, which was applied to French biotechnology firms, was by linking innovation with firm performance through the efficiency and efficacy of innovation performance [66]. The efficiency of an innovation reflects the degree of success of the innovation while efficacy indicates the effort carried out to achieve that degree of success [66]. Efficacy and efficiency, considered as complimentary dimensions that shape innovation performance, are measured through twelve items, including market share, new markets, cost per innovation, average number of innovation projects, working hours and product range extension [66].

Literature-based Innovation Output (LBIO) data has also become increasingly popular as a means of measuring innovation. LBIO data is compiled by screening specialist trade journals for new-product announcements instead of drawing on R&D figures that are seen as being not comprehensive [67]. Such methods have been applied to public service innovations

as well as product innovations [68]. LBIO data methods correctly compiled have no biases, are cost effective and can be an alternative to other innovation data, though they do tend to over-estimate domestic innovation and are limited where there are relatively few trade journals in a specific industry [67].

Research and Development data has been put forth as an innovation measurement. R&D measurement is seen by some as a poor measure since many small companies are innovative but spend little on R&D [69]. This data is also an input data, showing what was spent to get to a possible innovation and not indicative as an output of implemented innovations. Other measurement methods include valuation by royalties [70], radicalness and relative advantage [71], and radical versus incremental innovation [72].

Measurement with Patents. Though many options are presented concerning measurement of innovation without the use of patent data, the majority of the literature presents methods of measurement that are based upon some form of patent data. As indicators of technological change or innovation, patents have several advantages. Some of their advantages are:

- They are a direct outcome of the inventive process, and more specifically of those inventions, which are expected to have commercial impact. They are a particularly appropriate indicator for capturing the proprietary and competitive dimension of technological change.
- Because obtaining patent protection is time-consuming and costly, it is likely that applications are filed for those inventions, which, on average, are expected to provide benefits that outweigh these costs.
- Patents are broken down by technical fields and thus provide information not only on the rate of inventive activity, but also on its direction.
- Patent statistics are available in large numbers and for a very long time series.
- Patents are public documents. All information, including patentees' names, is not covered by statistical confidentiality. [64]

Patent data is easily accessible and cost-free through many national and international patent offices such as the United States Patent and Trademark Office (USPTO) and the

European Patent Office (EPO). The ease and simplicity of acquiring patent data make the use of it an obvious choice for analyzing of invention and innovation.

Patents are often cited as indicators of innovative growth, however a rigorous study has never previously been carried out to determine if they are leading or lagging indicators. A study on urban and regional innovation in metropolitan statistical areas (MSA) has found that in the absence of a better set of indicators that patents can serve as a rough measure of innovation. Further, because inventors frequently seek patent protection for new knowledge or processes, patents can serve as a proxy for innovation [73]. A metropolitan area's innovative strengths and growth rates can be indicated through patent data by technical classification of the patents [73]. Patent analysis can provide assistance in strategic planning efforts to firms involved in the ITS (Intelligent Transportation Systems) sector. ITS covers the application of computer, communication, positioning, sensing, control and other systems used to improve aspects of surface transportation. Patent information from the USPTO, EPO and JPO (Japanese Patent Office) concerning ITS and other related worldwide patent developments has been used to assess and provide an overall picture of ITS innovations and future markets [74].

Examples of innovation measurement techniques that are not technology or business specific but in some way depend on patent data information are available from various sources [76-77]. The patent success ratio (PSR) is defined as the ratio of successful patent applications to total patent applications. Supporters of this method claim that the PSR is an accurate measure of how innovative activity has changed over time. Correlations between the PSR and economic growth, or gross domestic product (GDP) are often claimed as being better than the correlation between successful patents and GDP, thus making PSR a better proxy for innovation [75]. The citations made in any patent document have been suggested as indicators of innovation, knowledge flows and spillovers, and thus of technological impact.

The suggestion is often made that the importance and impact of patents are greater when they are cited in succeeding patents [77,78]. The preceding discussion illustrates the divergence in opinion concerning the measurement of innovation. Much time and effort has been spent in seeking the discovery of an accurate measuring technique of something that itself is difficult to perfectly define. On balance the available literature appears to favor the use of some form of patent data as an indicator for innovation and innovative activity.

This dissertation uses patent data as an indicator of innovation and introduces a new long life cycle approach to qualify patents when they become important as innovation drivers. Short-term product cycles have been well studied [3-33]. Long-term life cycles, however, are relatively unstudied [2]. Long-term life cycles are possibly more suited to explain the impact of patents on production activity.

Section 2: Data Collection

Engineering Material Production Activity Data Collection. Production activity data was collected from the United States Geologic Survey (USGS) web site [78]. In all fifty cases, world production, by year, of the material in question was used for the activity data. This information was found in the historical statistics compilations of the minerals section of the USGS web site [79]. All activity is reported in metric tons. World production was chosen as the basis for activity due to its relevance to activity in global materials production, the completeness of the data sets available and the generally comparable definitions of world production between the individual materials. World production, for the most part, is based upon primary mine production with specific production definitions for each material being available in Appendix 3. The inclusion of recycled materials, beyond those generally included in primary production, was decided against due to the unavailability of consistent worldwide recycling data covering the years researched. Also, where recycling production

data was available it was apparent that the amount of world production attributed to recycling was at times already included in the primary production numbers or not significant to the world production totals.

Patent Data Collection. Data for patents published per year was collected from the European Patent Office (EPO) using its patent search engine [80]. The EPO web site was chosen because it offered the widest database for collection of global patents from 1900 to the present, while the United States Patent and Trademark Office (USPTO) on-line patent data only goes back to the 1970's. The EPO site provides worldwide searches, encompassing the patent offices of over 80 countries and regional intellectual property organizations including the United States, Germany, Great Britain, Japan, Korea, India, China, the EPO and the World Intellectual Property Organization (WIPO) [80]. The use of the EPO worldwide search, which includes countries who are signatories of the Patent Cooperation Treaty (PCT), is thus more thorough in coverage and scope than searches used to evaluate patent counts and innovation as presented, for example, by reference [74] above in section 1 which uses patent counts to investigate future markets and innovation.

The patent search was conducted through the EPO, using keywords to be found in the title or the abstract of the patent by the year of publication of the patent. As examples, for aluminum, the keywords employed were aluminum, Al and aluminium and for zinc, the keywords chosen were zinc and Zn. All other materials were done in a similar manner.¹ Multiple patent counts of the same patent, caused by multiple filings of patents in different patent offices, were not an issue as the search produced only one listing per patent. This listing would have the various patent offices through which the patent application was published, but would only list it once in the yearly patent count.

¹ A complete list of patent search keywords is included in Appendix 6.

Title and abstract fields were chosen, as it was clear that they provided the most likely option for finding the most complete set of patents. Also, this choice was made because the EPO search engine does not provide a claims field, which would have been preferred. Choosing the titles field gives any patent with the selected keywords in the title of the patent document. The abstract field will indicate any patent with the keywords in the abstract and gives any patents that do not include the keywords in the title, which is often the case.

The date of publication field was opted for as well. This selection provided the most complete set of data concerning patents containing the keywords and also was made due to the fact that there was no field for the date of patent issuance. The patents listed in this search field were not necessarily granted patents, but in some cases may be applications that are still pending. The date of publication, is not necessarily the date of issuance, but exhibits an accurate model of the relative innovation occurring during a specific year, since the existence of innovations is proven by publication rather than by an issuance of a patent for them.

Section 3: Patent and Production Activity Data Correlation

This section describes the method of comparison between the data gathered, representing the production activity per year of an engineering material, and the number of patents published per year for the same material. In correlation theory, two data sets, x and y , are tested to determine the existence of correlation between them. In this case, x is the production activity of a specific material in metric tons per year and y represents the number of patents published involving the same material for the same year. Through correlation theory, a number called the correlation coefficient is generated that expresses the amount of correlation that exists between two groups of data [81-83]. When the coefficient is squared and then multiplied by 100 a percentage is given, which expresses the percentage that changes in group y can be attributed to changes in group x [81,82]. In this manner, a

percentage of the changes in the patent numbers of a material were attributed to changes in the production of the material [83]. A more detailed description of the calculation of the correlation coefficient with an example can be found in Appendix 1. Examples of strong and weak correlations are discussed below in Figs. 1 and 2. A graphical representation of the correlation between the rare earth elements activity and patents is presented in Fig. 1, which indicates that the curves for activity and patents have a strong correlation to each other. A weak correlation is shown in Fig. 2, for beryllium.

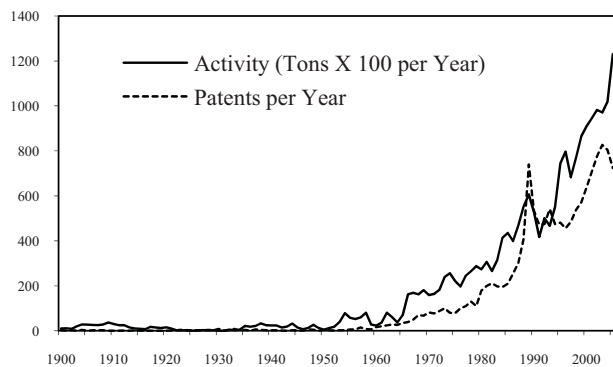


Figure 1. Rare Earths: Activity and Patents. A strong correlation is illustrated by this figure. The curves track each other fairly well and much of the change in the patents can be attributed to changes in the production according to correlation theory. The calculated correlation coefficient was 0.9659. Data scaled to fit on same figure.

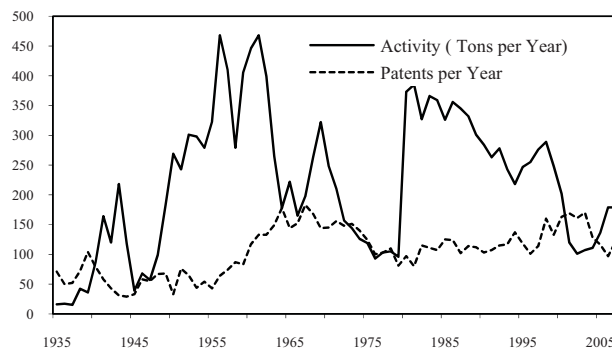


Figure 2. Beryllium: Activity and Patents. Weak correlation is exhibited in this figure with the two curves showing scant resemblance. Even visually it is noted that a change in the patents cannot be attributed to change in the production and vice-versa. The calculated correlation coefficient was 0.1132. Data scaled to fit on same figure.

The evaluation presented here indicates weak to strong relationships existing between the material activity and patent data sets depending on the material under study. As is shown below in Table 3, most materials investigated showed some degree of correlation between their activity and their patents. Table 3 gives comparative results arrived at after application of the correlation equations to the remaining forty-eight materials in the same manner as the previous examples for the rare earths and beryllium.

Table 3. Overall Correlation Coefficients (r) and $100r^2$. These r values represent the percent of variations in one data set, that affect variations in the other set for all materials studied. The best correlation is when r is one (i.e. 100%). An $(100)r^2$ of 90% means that 90% of the differences between points in one set of data can be attributed to corresponding differences in the other set of data [81-83].

Material	Overall Correlation Coefficient (r)	(100) r^2	Material	Overall Correlation Coefficient (r)	(100) r^2
Aluminum	0.9652	93.16%	Magnesite	0.8108	65.74%
Antimony	0.8518	72.56%	Magnesium	0.9078	82.41%
Arsenic	0.3375	11.39%	Manganese	0.6835	46.72%
Asbestos	0.7288	53.11%	Mercury	-0.1117	1.25%
Barite	0.7486	56.04%	Molybdenum	0.9290	86.30%
Bauxite/Alumina	0.9310	86.68%	Nickel	0.9563	91.45%
Beryllium	0.1132	1.28%	Niobium	0.7543	56.93%
Bismuth	0.7651	58.54%	Nitrogen	0.9164	83.98%
Boron	0.8691	75.53%	Phosphate	0.8708	75.83%
Cadmium	0.7401	54.77%	Platinum	0.9569	91.57%
Chromium	0.9495	90.16%	Potash	-0.1414	2.00%
Cobalt	0.9269	85.91%	Rare Earths	0.9700	94.09%
Copper	0.9507	90.38%	Salt	0.8996	80.93%
Feldspar	0.9490	90.06%	Selenium	0.6871	47.21%
Fluorspar	0.7399	54.75%	Silicon	0.8893	79.09%
Gold	0.9385	88.08%	Silver	0.8735	76.30%
Graphite	0.9287	86.25%	Sulfur	0.9156	83.83%
Gypsum	0.9211	84.84%	Talc	0.9383	88.04%
Helium	0.7460	55.65%	Tantalum	0.7181	51.57%
Hydraulic Cement	0.9299	86.47%	Tin	0.6576	43.24%
Iodine	0.9105	82.90%	Titanium	0.9151	83.74%
Iron	0.8741	76.41%	Tungsten	0.7420	55.06%
Kyanite	0.9242	85.41%	Vanadium	0.7840	61.47%
Lead	0.7013	49.18%	Zinc	0.9387	88.12%
Lithium	0.9272	85.70%	Zirconium	0.9239	85.36%

The summary presented by Table 3 indicates that forty-eight of fifty materials investigated, according to the correlation methods referred to, establish a possible

relationship between the material's activity and patent data. Most of the materials tested showed some degree of correlation, although correlation for arsenic and beryllium was weak and mercury and potash had negative correlation, which indicates a lack of a relationship. Correlation theory has thus shown that the data sets of production activity and patents for the materials evaluated here are not randomly connected, but are in fact related to each other. This relationship implies that a change in patent trends is due to a corresponding change in the production activity allowing for the confident use of these data sets in further evaluations employing best-fit models of selected metals and non-metals.

From the discussion above it can be concluded that there is a correlation between the data gathered concerning material activity and the numbers of patents published, which represent innovations utilizing such materials, making further comparisons and evaluations of the data sets more valid. This has been proven employing standard statistical procedures for forty-eight of fifty materials studied in this paper. The variations in the patent data during the dominant part of the life-cycle, namely Stage 3, illustrated in Figure 3 below, [2,3] do not appear to occur on their own, but can be correlated to the variations in the material production activity.

Where correlation exists, as here in reference to engineering material production and patenting data, the disclosed relationship could possibly be used for predicting the future behavior of one set of data based upon knowledge of future behavior of the other data set. For example, if a material has strong correlation between production and patenting, or innovation, and the government announces that it will provide billions of dollars for innovative development of the material, it might be a good option to provide resources for future production of the material since correlation theory predicts a rise in production will mirror the announced increase in innovative activity measured by patents.

Section 4: Best-Fit

Now that the correlation between activity and patents has been established, these sets of data can be used in conjunction with the common pattern equation for production of metals initially proposed by Yerramilli and Sekhar [2] and modified by Connelly and Sekhar [1]. A further modification is proposed below in this section. The equation predicts and illustrates a four-stage life cycle for metals. These four stages are the Initial Stage (I), the Lift Off and Decay Stage (II), the Revival and Rapid Growth Stage (III) and the Survival Stage (IV) [2]. The patterns found are common to the materials tested and are similar to common patterns and cycles found in overall life behaviors as illustrated by long wave theory [2]. A similar hypothesis for shorter life products, which postulates that that most successful products pass through recognizable stages during their life cycles was first proposed by Levitt and has been applied by others in evaluations of industry and business activities for various products and product groups [84-89].

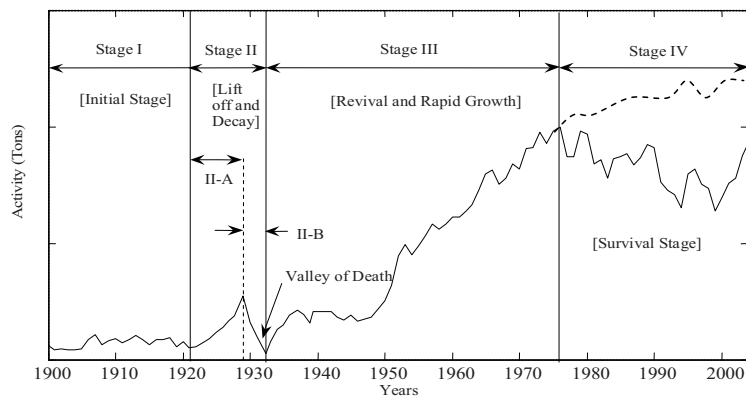


Figure 3. Illustration of a Typical Long-term Life Cycle for a Metal from References 1-3. The plot indicates the division of the life cycle into four stages that is common to metals. All metals may not have all four stages depending on the length of time that the metal has been in use. Such life cycles are also applicable to non-metals.

Figure 3 illustrates an example of a life cycle of a material, comprised of the four previously mentioned stages. Stage I (*Initial Stage*) is the developmental stage that begins with the discovery and the invention of a process and ends when the development of the

technology is enough to start low-scale industrial production of the material. Stage II (*Lift Off and Decay Stage*) begins with the rise in the activity of the material and ends at the low point of the activity in the so-called “valley of death.” Stage III (*Revival and Rapid Growth*) begins at the “valley of death” and continues through the material’s full growth potential with the take-off in activity typically being at a high rate. Stage III ends at the onset of Stage IV (*Survival or Low Growth Stage*) where the material has reached maturity and the activity has leveled off or has begun to die [2,3]. Invention driven activity occurs during Stages I and II while innovation is dominant in Stages III and stage IV. Stages I and II are the periods, in the life of a material, where the invention itself is developed into a market innovation with technological R&D, possibly being very important. Stages I and II may be technically driven whereas Stages III and IV include market, teaming and financial factors. The type of leadership required may evolve as the stages are transitioned [1-3]. Dramatic changes in Patents and Production activity occur in Stages III and IV compared to the early stages, i.e. when innovation becomes the major focus. In these latter stages any early invention is fully developed, mature and possibly patented. R&D is complete for the most part and marketplace interest in the product begins to develop. Stages III and IV are the times in the life cycle of the product for commercial exploitation. Market place effectiveness as well as cost and profit issues predominate. If the invention has not previously been patented, it may be patented now as a means to protect the invention as well as any follow-on innovation that arises from it [1].

Platform equation with Chromium as an example. As shown in Section 3, chromium displays good positive correlation between its activity and patent data. Chromium has been widely used for over several decades. This wide use has provided strong and consistent numbers for production activity as well as patents per year.

The best-fit method requires the determination, by trial and error, of multiple parameters to be entered into an equation and a MatLab computer program, examples of which are found in Appendix 3. The common pattern equation for this method is

$$y = x^n [\alpha^n x^2 + \beta^n x \sin(\omega x)] + (\exp[(x - \mu) / v] \exp[-\exp[(x - \mu) / v]]) \delta / v, \quad (1)$$

with the variables, α , β , ω , n , μ , v and δ defined in Table 4 to be determined for each material tested. The original equation in reference [2] is modified by substitution of α and β with α^n and β^n to eliminate the possibility of multiple values of α and n giving equally acceptable R^2 . Only one set of α and n give an obvious best R^2 , which is the value consistently chosen, with the α^n and β^n equation while the α and β equation might give a less obvious choice concerning the generated R^2 value.

Table 4. Common Pattern Evaluation Variables. Variables to be determined in connection with the common pattern equation (1). Normalized years are the span of years under consideration and are represented by x . Production is y . The remaining variables are found through trial and error.

α	Called the “Take-off constant”. Facilitates the rate of take-off after the end of Stage II. The rate of growth of activity is very sensitive to α . Dimension is dependent on n .
β	Increases the amplitude (visibility) of the cyclicity. Magnitude of cyclicity increase as β decreases. The dimensions of β are dependent on n . Dimension is dependent on n .
ω	Called the “wavelength constant”. Increases in ω increase cyclicity. Value of ω expressed in “per year” and equals $(2*\pi)/\text{wavelength}$.
μ	Called the “Stage II location constant”. Position of the Stage II hump is shifted to the right as value of μ increases and is expressed in “years”.
v	Called the “Stage II scaling constant”. As value of v increases, the Stage II hump is stretched out and is also expressed in “years”.
δ	As the value of δ increases, the peak (amplitude) of the Stage II hump increases. δ is given in tons.
n	Along with α has a strong influence on the shape of the curve. It is a positive number between 0 and 2. n is dimensionless. Dimensionlessness is assured through normalizing by dividing by n_0 , which is one.
x	Time in normalized years. Actual year of data (x_r) minus year of origin (x_0).
y	Metric tons per year (In some cases scaled to kilotons, megatons or kilograms). Entered by thousands.

These seven parameters, as well as the date of origin, x_0 , of the data are entered into a MatLab computer program which generates an actual curve of the data, a fitted curve and an

R^2 value, which is an established measure of best-fit, and which needs to be as near one as possible to obtain the best fitted curve [1]. The origin, x_0 is determined from the earliest available production data and is usually 1900. Some materials have an x_0 later than 1900 due to lack of data back to 1900 because of missing data or a material whose production, or use, did not begin until after 1900.

Equation 1, from reference [1], is a modified form of the first published form of equation 1, in references [2,3]. It is further noted that n is non-dimensional while α and β have the same units as y . For $x=(x_r-x_0)$, x_r is the actual year of the data and x_0 is the first year of the data set. Also, $n=n/n_0$ where n_0 always takes the value of 1 (below it is noted that n_0 appears to be a universal constant which is determined from Figures 12(a-c)).

The optimal parameters and resulting R^2 value for chromium activity are listed below in Fig. 4. It has been determined that α and n cause the most drastic change in R^2 and are used during the fitting to get the R^2 value closest to one. The origin, x_0 , is simply the first year of the data. For chromium, an R^2 of .9731 was found for its production data, which produces a best-fit curve that tracks, the actual data curve that displays Stage III features and indicates a Stage III best-fit for chromium production.

In general, as with chromium, the production is entered in the best-fit production equation as thousands of tons. For many materials this leads to a high R^2 value and also to an eventual shift in origin. In other cases the production must be scaled differently to achieve a high R^2 value and an origin shift. Scaling is sought to achieve the least differential between the scale of the production and patent data. A material with production numbers much larger than its patent numbers may need to be scaled up and entered as thousands of kilotons or megatons. When patent data counts are much greater than production, the production is entered as thousands of kilograms. This procedure allows for a more accurate evaluation of the best-fit of both production and patent data by resulting in R^2 values generally closer to

one and resulting origin shifts, but also is representative of the same amount of production only in a different scale. It is possible that the relative changes from year to year in the production data create the features in the plots that determine the stage of the material rather than the scale of the data. A material would be in the same stage whether its data is in tens, hundreds or thousands because the relative changes between data points would be the same from year to year resulting in identical plots with different y-axis scales.²

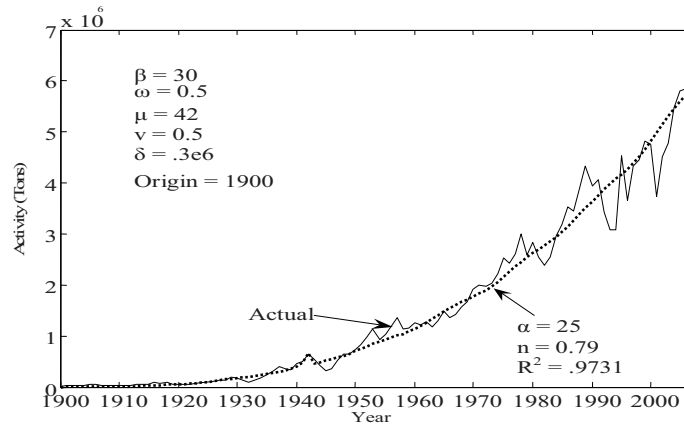


Figure 4. USGS World Chromium Production. Fitted chromium production activity curve with best-fit parameters and R^2 value and origin. The data is from the US Geologic Survey world chromium production by year [78]. Shown in the figure are both the actual data curve and a best-fit curve.

An identical best-fit equation was employed for the patent data using the same parameters as were found for the production activity to generate a modified patent fit curve as apposed to an independent patent best-fit curve which has parameters independent of the production common pattern equation. Minimal changes were made in the program code to allow for differences of the scaling for the plots, i.e. between the production activity and the patent data. The only changes in the parameters were the use of the number of patents data rather than metric tons of production for y , and the choice of origin, x_0 . Identical δ were used for modified best fits, but were scaled down on the figures, such as Fig. 5, by a factor of one

² A discussion of scaling is found in Appendix 7.

thousand to better reflect the scale of the Stage II hump in reference to the patent data. The patent data was not entered as thousands as was the production data, but was instead accounted for by scaling the δ on the modified best-fit figures. The origin for the patent best-fit was moved a number of years backwards or forwards relative to the origin of the activity best-fit equation, i.e. 1900 for chromium. An R^2 as close to one as possible was sought. In the case of chromium, as displayed on Fig. 5, a shift of origin to 1897, with all other parameters the same, gives an R^2 of .9320 which results in a fitted curve that tracks the actual patent data curve, having stage III attributes, and signifies a possible Stage III fit for chromium patent data.

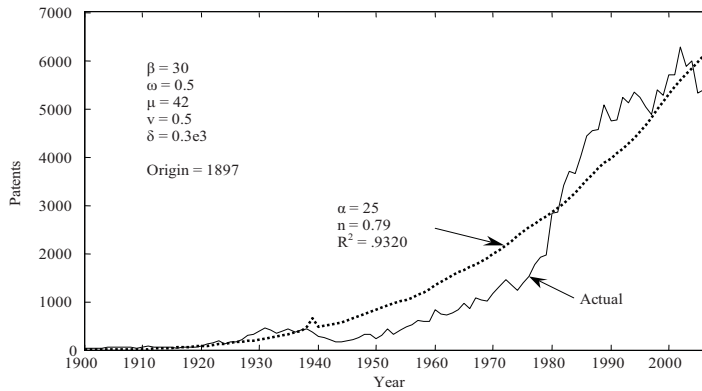


Figure 5. EPO Worldwide Patent Search: Chromium, Cr or Chrome in Title or Abstract by Date of Publication. Chromium modified patent fit curve with best-fit parameters and R^2 value and origin. The data is from the European Patent Office worldwide patents containing chromium or Cr in the title or abstract of the patent by date of publication. δ scaled down from .3e6 to .3e3 (See Appendix 7).

The origin of 1897 for the chromium patent best-fit curve signifies a 3-year lag for the activity data when compared to patenting. The 3-year lag of production activity can be observed by examining the Stage II hump of the activity curve and that of the patents. Fig. 4 shows the hump at roughly 1942 for the patent best-fit while Fig. 5 shows the hump for activity best-fit at about 1939. Three years after the patent data met this point in its life cycle, the activity data crossed the same normalized position in its own life cycle.

Table 5. Engineering Material Production, Independent Patent and Modified Patent R^2 Values, Correlation Coefficients (r), Origins, Origin Shifts and Stage. Materials are listed in order of descending Production R^2 values. An origin shift indicates the presence of a material in Stage III of its life cycle. Strong correlation and R^2 values near one are indicators of possible overall Stage III, but are not definitive evidence. Negative Production R^2 values, no correlation and lack of origin shift are indicative of Stage IV materials. (*Estimations were made to fill in gaps in the USGS production data for lead following patterns suggested by existing data. ⁺ Indicates materials using Equation (2) presented below). Possible Stage V materials are indicated.

Material	Production R^2	Independent Patent R^2	Modified Patent R^2	Correlation r	Production Origin	Patent Origin	Origin Shift	Stage
Nickel	0.9823	0.9502	0.7331	0.9563	1900	1831	(-)69 years	III
Aluminum	0.9818	0.9554	0.9658	0.9652	1900	1915	(+)15 years	III
Hydraulic Cement	0.9743	0.9328	0.9280	0.9299	1926	1925	(-)1 year	III
Chromium	0.9731	0.9368	0.9320	0.9495	1900	1897	(-)3 years	III
Copper	0.9576	0.9416	0.9320	0.9507	1900	1911	(+)11 years	III
Molybdenum	0.9538	0.9388	0.4214	0.929	1900	1712	(-)188 years	III
Platinum	0.9539	0.9121	0.6065	0.9569	1900	1810	(-)90 years	III
Bauxite/Alumina	0.9521	0.9074	0.4465	0.931	1900	1721	(-)179 years	III
Sulfur	0.9322	0.9678	0.2273	0.9156	1900	1650	(-)250 years	III
Talc	0.9226	0.9108	0.2900	0.9383	1904	1571	(-)333 years	III
Phosphate	0.8815	0.9668	0.5014	0.8708	1900	1741	(-)159 years	III
Zinc	0.8805	0.9617	0.9669	0.9387	1900	1918	(+)18 years	III
Cobalt	0.8796	0.9652	0.3584	0.9269	1901	1645	(-)256 years	III
Gypsum	0.8740	0.9543	0.7123	0.9211	1924	1865	(-)59 years	III
Lithium	0.867	0.9284	0.5000	0.9272	1925	1819	(-)106 years	III
Titanium	0.86	0.9620	0.9630	0.9151	1925	1926	(+)1 year	III
Iron	0.8599	0.9329	0.6888	0.8741	1904	1815	(-)89 years	III
Salt	0.8438	0.9782	0.3468	0.8996	1913	1663	(-)250 years	III
Kyanite	0.8387	0.9040	0.8825	0.9242	1928	1948	(+)20 years	III
Rare Earths	0.8256	0.8152	0.4483	0.97	1900	1799	(-)101 years	III
Niobium	0.8235	0.5007	0.5237	0.7545	1964	1827	(-)137 years	III
Magnesite	0.8231	0.7719	0.5681	0.8108	1913	1743	(-)170 years	III
Graphite	0.8122	0.9617	0.9045	0.9287	1900	1876	(-)24 years	III
Vanadium	0.8038	0.5164	0.7590	0.784	1960	1870	(-)90 years	III
Feldspar	0.7833	0.7775	0.4073	0.949	1908	1800	(-)108 years	III
Barite	0.7803	0.8540	0.3185	0.7486	1919	1760	(-)159 years	III
Magnesium	0.7502	0.9733	0.7010	0.9078	1937	1854	(-)83 years	III
Nitrogen	0.7316	0.8936	0.4491	0.9164	1946	1770	(-)176 years	III
Antimony	0.7192	0.9664	0.7644	0.8518	1900	1825	(-)75 years	III
Iodine ⁺	0.7185	0.3816	0.6347	0.9105	1960	1827	(-)133 years	III
Zirconium	0.6913	0.8964	0.9536	0.9239	1944	1923	(-)21 years	III

Potash	0.6511	0.0208	-	-0.1414	1951	-	No Shift	IV
Tungsten	0.6449	0.8969	0.3494	0.742	1905	1673	(-)232 years	III
Helium	0.6286	0.8457	0.4755	0.746	1935	1779	(-)156 years	III
Silicon ⁺	0.6029	0.6195	0.8980	0.8893	1964	1934	(-)30 years	III
Manganese	0.5728	0.9132	0.9666	0.6835	1900	1923	(+)23 years	III
Fluorspar	0.5703	0.8372	0.4717	0.7399	1913	1800	(-)113 years	III
Tantalum	0.5484	0.3886	-	0.7181	1969	-	No Shift	IV
Lead* ⁺	0.5368	0.9172	0.7303	0.7013	1900	1859	(-)41 years	III
Silver	0.5027	0.7582	0.3484	0.8735	1900	1921	(+)21 years	III
Boron	0.4209	0.4225	-	0.8691	1964	-	No Shift	IV
Asbestos	0.3850	0.0131	-	0.7288	1900	-	No Shift	IV-V
Gold	Negative	Negative	-	0.9385	1900	-	No Shift	IV
Arsenic	Negative	0.8399	-	0.3375	1910	-	No Shift	IV
Beryllium	Negative	0.0455	-	0.1132	1935	-	No Shift	IV-V
Bismuth	Negative	0.9463	-	0.7651	1937	-	No Shift	IV
Cadmium	Negative	0.8552	-	0.7401	1900	-	No Shift	IV
Mercury	Negative	0.8454	-	-0.1117	1900	-	No Shift	IV-V
Selenium	Negative	0.7969	-	0.6871	1938	-	No Shift	IV
Tin	Negative	0.9400	-	0.6576	1905	-	No Shift	IV

The remaining materials were also evaluated using the same best-fit procedure. However, in the cases of iodine, lead and silicon, where the first year of the production data is much greater than zero, a further modification was made to Equation 1 in order to obtain a more reasonable R^2 value. A constant, CI , was added to both the production and modified patent equation giving

$$y = CI + x^n [d^n x^2 + \beta^n x \sin(\alpha x)] + (\exp[(x - \mu) / v] \exp[-\exp[(x - \mu) / v]] \delta / v). \quad (2)$$

CI was equal to the first year of production for the specific material and allowed the fitted curve to match up better with the actual production producing an R^2 closer to one. In effect, the addition of CI is a form of scaling similar to that discussed above, and in Appendix 7, causing the actual production data to be closer to the generated fitted data.

Table 5 shows that seven materials had lags in their patent life cycles, indicating positive shifts forward in years in their fitted patent life cycles compared to their production activity life cycles. Positive shifts forward in origin mean that activity occurs before the patents and that patent output may be driven by the activity. Thirty-one materials have lags in their activity life cycles, rather than a lag in the patent life cycles, which illustrate negative shifts backwards in these materials' fitted patent life cycles. Negative origin shifts indicate that the patent production precedes the activity of the material and that the patents may drive the activity. Fig. 6 and Fig. 7 display these shifts graphically for chromium and zinc and show a negative and positive lag respectively.

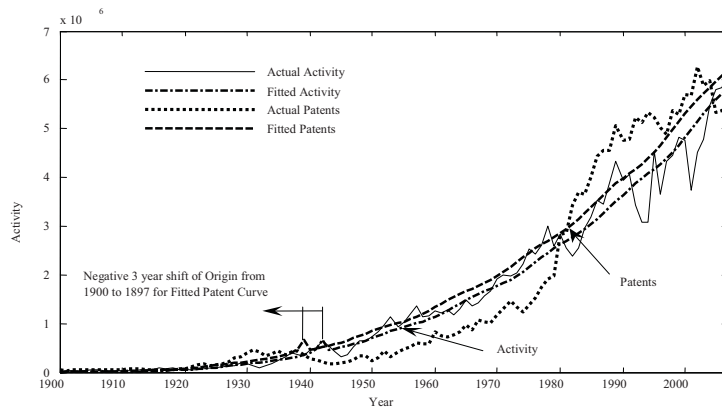


Figure 6. Chromium Best-Fit Activity and Patents. Plot showing the origin shift of patent and activity best-fit curves for chromium. The shift is negative, indicating patent activity occurring before production activity and thus possibly driving the production.

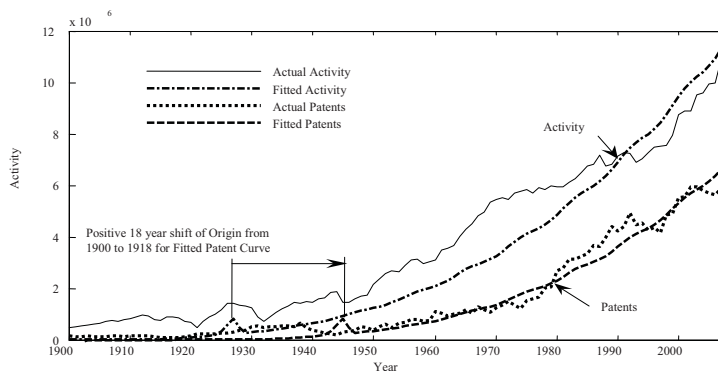


Figure 7. Zinc Best-Fit Activity and Patents. Plot depicting the origin shift of patent and activity best-fit curves for zinc. The shift is positive, indicating patent activity occurring after production activity and thus possibly being driven by the production.

The shifts in the fitted patent life cycles appear to be dependent on the material itself and outside factors affecting production and patenting. A negative shift, or lag in the activity life cycle illustrates where patenting precedes the production of the material and suggests that patenting may drive production. In such cases the patent represents the innovation that drives the economy and causes production of the material. The positive shift, or lag, of the patent life cycle may be attributed to the case where production of a material precedes the patenting of ideas related to that material. Invention and innovation follow and are possibly driven by production of the material. A lack of a shift occurred in all cases where Stage IV behavior might be evident and may indicate a lack of innovation [1,83]. The case where patents drive activity could be analogous to innovation driving the economy in a creative manner. Patents give an incentive to innovate by offering property rights and cause increased production activity as a result. In the same way, the destructive activity of innovations, where they destroy to build the economy anew, may be analogous to activity leading patents, where patents are employed to prevent innovators from effectively competing [1,83].

Correlation theory and best-fit analysis provide tools for the examination of the life cycle of a material. Through study and manipulation of the production and patent data of materials, curves can be generated that can be used as indicators in determining the stage where a specific material resides in its life cycle by looking for the classic features common to the four identified material stages identified in Fig. 3.

Stage Indication

The estimated stage of the material is based upon a combination of indicators that point towards the stage in its life cycle that the material is in. Strong correlation is the first indicator of possible Stage III (Table 3), which is illustrated by a graphical plot of the production and patent data. At times the plots relating to correlation reveal curves with Stage

III features. This first indicator signals possible Stage III when the correlation coefficient is strong, or approaching one. Best-fit R^2 values approaching one for either or both the production activity and patent data of a material are also indicators of a possible Stage III material. Such R^2 values are produced for curves that have Stage III life cycle attributes as shown in Fig. 3. and are generated from Eq. 1 using linear alphas, which are found to generate Stage III best-fit curves [2].

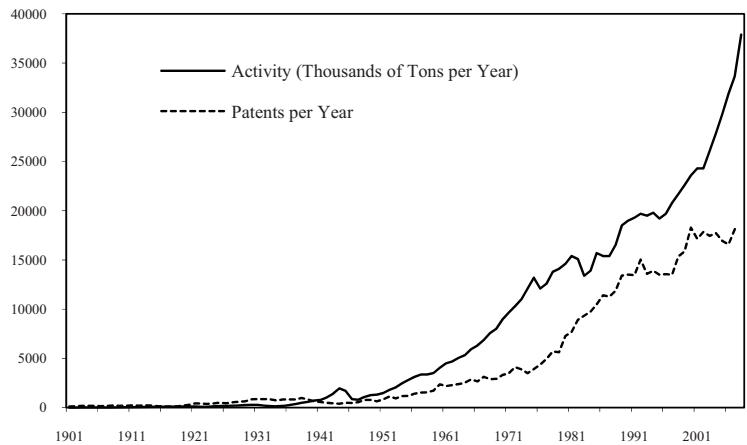


Figure 8. Aluminum: Activity and Patents. Aluminum Production curve displaying typical Stage III features. “Valley of Death” crossed in 1946 with high rate of growth since and into 2007 with patents tracking the growth, indicating continuing innovation.

Aluminum is shown to be an estimated Stage III material in Fig. 8, where in 2007 it is still exhibiting rapid innovative stage (stage III) growth and has strong correlation and R^2 values approaching one. Stage IV behavior is displayed by arsenic in Fig. 9. Arsenic shows no rapid growth, but instead a general leveling off with local oscillations depicted, indicating less innovative activity and has weak correlation and a negative production R^2 . In examples such as aluminum and arsenic the stage of the material is suggested by examining the plots of the activity and patents as well. In cases where there is no clear delineation between stage III and IV, as in manganese in Fig. 10, the correlation coefficient, r , and the best-fit R^2 and

origin shifts can be employed as indicators in the determination of the possible stage of the material.

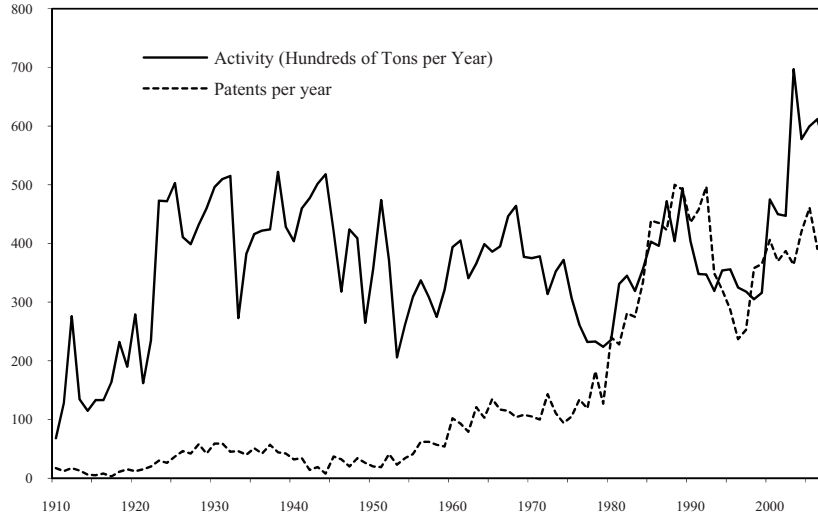


Figure 9. Arsenic: Activity and Patents. Arsenic production curve displaying typical Stage IV features. This material appears to have reached maturity with no more sustained rapid growth and activity generally leveling off. Oscillations for growth and shrinkage are common in Stage IV materials. Production is stagnant in a time averaged since, but patents grow in number.

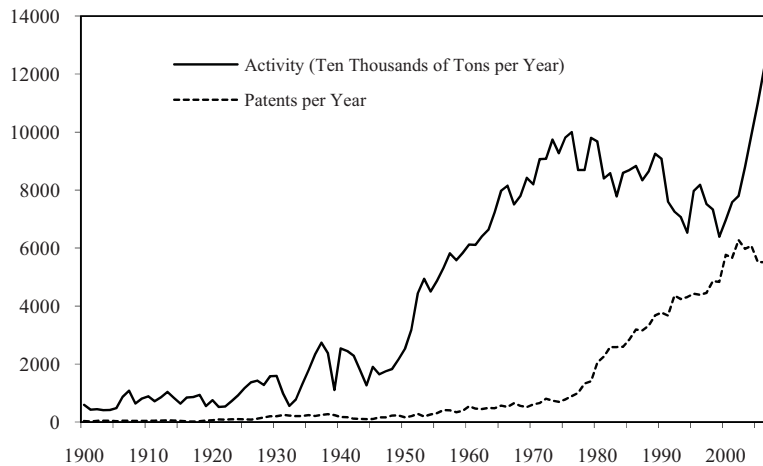


Figure 10. Manganese: Activity and Patents. Manganese production curve possibly illustrating a material fluctuating between Stage III and IV. As indicated by Best-fit analysis below Manganese is in Stage III in 2007. However, in 2005 when the data was studied, Mn appeared to be a Stage IV material. This suggests that materials can be in a place in their life cycle where they fluctuate between stages depending on the production numbers for those years.

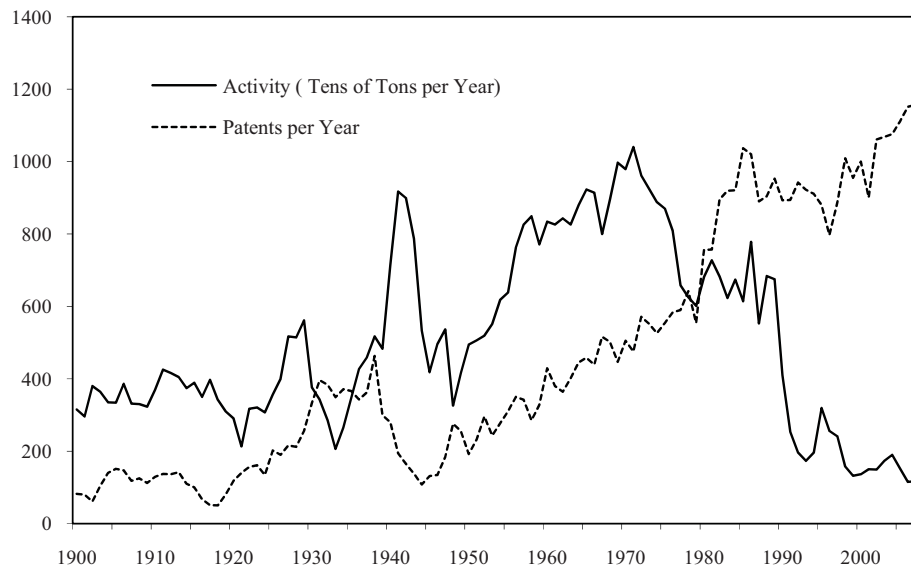


Figure 11. Mercury: Activity and Patents. Mercury production curve displaying a possible Stage V. Production is well beyond the survival stage and has greatly dropped in a sustained manner. Correlation theory indicates little relationship between production and activity and Best-fit analysis produces a negative R^2 . There is very little production for any patents to drive.

Mercury, in Fig. 11, presents an interesting case. It has a negative R^2 , its correlation coefficient is negative, which indicates no relationship between production and patents, and there is no shift in its origin. Fig. 11 presents a curve that reveals no growth, and not even a leveling off, but instead a sustained downturn in production having no signs of stopping. Mercury appears to be in its death throes of production, due to its toxicity and environmental concerns regarding it [90]. Production of this material is being replaced by recycling of presently available material. Any innovation associated with mercury is generally innovation away from it or in replacement of it, which would explain increasing patenting activity seen in Fig. 11 [90]. Such a material could be in a Stage V or “Final Death Stage.” Other toxic materials such as asbestos and beryllium exhibit such behaviors and could be called Stage V materials as well.

Sometimes the stage of a material is suggested by the appearance of its curves. In other cases the stage is not obvious, but the combination of these indicators allows an intelligent inference of the stage of a metal or non-metal. From this, assumptions can be made concerning the innovative activity, past and present, of a material allowing for more educated and informed decisions concerning the future behavior of a material and more efficient development and use of them. A more definitive method is needed that will strongly identify a material as being in Stage III rather than just indicating the possibility of it existing. Such a novel method is presented below utilizing the origin shift of the patent best-fit data that is present in all materials that exhibit the above mentioned Stage III indicators.

Section 5: Best-Fit, Origin Shift and Innovation

The best-fit approach can be applied to comparisons of activity and patent data with patent and origin shifts, allowing inferences to be made concerning the relationship of innovations to production. For this application, the best-fit equation and program were applied to all of the studied materials' activity and patent data independently. This analysis was performed successfully on the materials listed below in Table 6. These materials were estimated to be in Stage III since, in general, they had strong correlation, High R^2 values and production plots that revealed common Stage III attributes. Origin shift analysis was not successful in the remaining materials where an origin shift could not be found and where such materials were considered to be in Stage IV due to their weak correlation, low R^2 values and production curves that exhibited Stage IV features. The presence or absence of an origin shift verified these stage assumptions in all cases.

Table 6. α and n Parameters, α^n Ratio and Origin Shifts, Origin Ratios and Modified R^2 .

Alpha and n are from the pattern equation. Ratio of α^n indicates strength of the driving force of the material. The farther the ratio is from one, in either direction, the greater the driving force. A positive origin shift could indicate patents being driven by production. A negative origin shift suggests production being driven by patents. A positive origin shift results in an origin ratio greater than one, while a negative origin shift leads to an origin ratio less than one. Note that (n_d/n_p) is less than one when the origin shift is negative and one or greater when the shift is positive. Modified R^2 generally becomes smaller than one as the origin ratio moves farther from one. The materials are listed by descending order of origin shift and origin ratio.

	α_a	n_a	α_a^n	α_p	n_p	α_p^n	Drive Ratio	n_d/n_p	Origin Shift	Origin Ratio	Modified R^2
Manganese	23	0.9	16.81	22	0.8	11.86	1.418	1.13	+23	1.012	0.5728
Silver	5	1.2	6.90	3	1.1	3.35	2.060	1.09	+21	1.011	0.3484
Kyanite	22	0.6	6.39	16	0.5	4	1.597	1.2	+20	1.010	0.8825
Zinc	20	0.9	14.82	24	0.8	12.71	1.17	1.13	+18	1.009	0.9669
Aluminum	13	1.1	16.80	15	1.0	15	1.12	1.1	+15	1.008	0.9658
Copper	13	1.0	13	9	1.0	9	1.44	1	+11	1.006	0.9320
Titanium	19	1	19	18	1	18	1.06	1	+1	1.001	0.9630
Hyd. Cement	15	0.5	3.87	15	0.51	3.98	0.973	0.980	-1	0.999	0.9280
Chromium	25	0.79	12.72	26	0.8	13.55	0.939	0.99	-3	0.998	0.9320
Zirconium	16	0.85	10.56	15	1	15	0.704	0.85	-21	0.989	0.9536
Graphite	18	0.6	5.66	16	0.7	6.94	0.813	0.86	-24	0.987	0.9045
Silicon	16	1.2	27.85	16	1.5	64	0.435	0.8	-30	0.985	0.8980
Lead	12	0.8	7.30	23	0.9	16.81	.434	0.89	-41	0.978	0.7303
Gypsum	28	0.4	3.79	34	0.6	8.30	0.457	0.67	-59	0.969	0.8740
Nickel	43	0.58	8.86	14	0.9	10.75	0.824	0.64	-69	0.964	0.7331
Antimony	42	0.3	3.07	35	0.5	5.92	0.519	0.6	-75	0.961	0.7644
Magnesium	15	0.7	6.66	25	1	25	0.266	0.7	-83	0.957	0.7010
Iron	13	0.7	6.02	27	0.9	19.41	0.31	0.78	-89	0.955	0.6888
Vanadium	75	0.4	5.62	21	0.9	15.49	0.363	0.44	-90	0.954	0.7590
Platinum	14	0.5	3.74	31	0.7	11.07	0.338	0.71	-90	0.953	0.6065
Rare Earths	22	0.26	2.23	33	0.5	5.74	0.389	0.52	-101	0.947	0.4483
Lithium	23	0.5	4.80	16	0.9	12.12	0.396	0.56	-106	0.945	0.5000
Feldspar	11	0.01	1.02	11	0.3	2.05	0.499	0.033	-108	0.943	0.4073
Fluorspar	1	0.003	1	15	0.3	2.25	0.444	0.01	-113	0.941	0.4717
Iodine	10	0.4	2.51	9	1	9	0.279	0.4	-133	0.932	0.6347
Niobium	16	0.5	4	13	1.16	19.6	0.204	0.43	-137	0.930	0.5237
Helium	20	0.23	1.99	15	0.7	6.66	0.299	0.33	-156	0.919	0.4755
Phosphate	30	0.35	3.29	27	0.7	10.05	0.327	0.5	-159	0.916	0.5014
Barite	14	0.017	1.05	18	0.4	3.18	0.329	0.04	-159	0.917	0.3185
Magnesite	35	0.08	1.33	43	0.4	4.50	0.295	0.2	-170	0.911	0.5681
Nitrogen	14	0.56	4.38	13	1.2	21.71	0.202	0.47	-176	0.910	0.1770
Alumina	59	0.3	3.40	26	0.7	9.78	0.347	0.43	-179	0.906	0.4465
Molybdenum	17	0.36	2.77	36	0.7	12.29	0.226	0.51	-188	0.901	0.4214
Tungsten	27	.23	2.13	17	.7	7.27	0.294	0.33	-232	0.878	0.3494
Sulfur	20	0.25	2.11	16	0.8	9.19	0.230	0.31	-250	0.868	0.2273
Salt	19	0.47	3.99	20	1	20	0.200	0.47	-250	0.869	0.3468
Cobalt	24	0.2	1.89	15	0.7	6.66	0.284	0.29	-256	0.865	0.3584
Talc	1	0.01	1	17	0.5	4.12	0.243	0.02	-333	0.825	0.2900

The use of the pattern equation creates a relationship between α and n that can be evaluated and compared to origin shifts produced by independent patent and production activity best-fit derivations. A graphical representation, such as Fig. 12, of the relative scale, or distance, of the origin shift can be made, using a ratio of the shift and the origin, x_0 , of the

production data, indicating an absolute amount that the patent or activity driving force has on the other. This ratio, called the Origin Ratio, composes the x-axis of Fig. 12 and is defined as

$$\text{Origin Ratio} = (x_0 + OS)/x_0 \quad (3)$$

where x_0 equals the production data origin and OS is the shift in origin of the best-fit patent data. The origin ratio is dimensionless since x_0 and OS are both in years which are then cancelled out. The y-axis of Fig. 12 is the drive ratio of the material and is expressed as

$$\text{Drive Ratio} = (\alpha^n)_a/(\alpha^n)_p \quad (4)$$

where $(\alpha^n)_a$ equals the modified patent best-fit variable alpha to the n power, which is the same as the alpha and n from the production activity best fit equation, and $(\alpha^n)_p$ is equal to the independent patent best-fit variable alpha raised to the power n , in both cases n , being best-fit variables. The drive ratio is dimensionless as well since $(\alpha^n)_p$ is generated from a best fit equation having data with patents as units. Likewise, $(\alpha^n)_a$ results from a modified patent equation, that is used to generate the origin shift, having units of patents. These units cancel each other upon calculation of the ratio. Such a curve with the origin ratio on the x-axis and the drive ratio on the y-axis may effectively represent innovative behavior.

Table 6 presents the α and n values for the independent activity and patent data and the original origin shifts and origin ratio derived for predicted Stage III materials in Section 4 as well as α^n and drive ratios. The drive ratio generally becomes progressively larger than one, as the origin ratio grows larger than one, which represents the origin shift moving away from zero in a positive shift direction. Likewise, the drive ratio approaches zero as the origin

ratio becomes progressively smaller away from one, which is representative of the origin shift moving further in the negative direction from zero as shown in Figs. 12(a, b and c) below. The figures show the calculated positions for all those materials that possess a shift in origin. Note that the activity of nitrogen is possibly being driven the most by its patents since its ratio is nearest to zero for materials whose activity is driven by patents. Similarly, the patents of silver may be driven the most by its activity because its ratio is farthest from one for materials whose patents are driven by activity.

Three patterns emerge from the best-fit analysis shown in Table 6. The first is that when a positive origin shift is indicated, the drive ratio is always above one (the ratio is always less than one and approaches zero when the origin shift is negative). Second, when n_a is divided by n_p the result is always less than one for materials that have negative shifts in origin and the result is always one or greater for materials with a positive origin shift. While n_a is the value for the n variable for the activity best-fit evaluation, n_p is the value for the independent patent best-fit evaluation. The ratio between them may be indicative of the driving force of the material. Lastly, as the origin ratios move away from one in either direction the modified patent R^2 , which is generated by patent data being run with the common pattern equation production parameters, generally becomes smaller than one. As indicated in Table 6 and on Fig. 12(a) materials such as talc, sulfur, cobalt and salt have origin ratios the farthest below one and lower modified R^2 than do materials with ratios nearer to one. In the same way silver, and manganese have the highest origin ratios and have R^2 's less than materials with origin ratios near to one. This could indicate that materials approach Stage IV when they reach origin ratio extremes and enter stage IV when modified R^2 values become too small to support an origin shift or origin ratio.

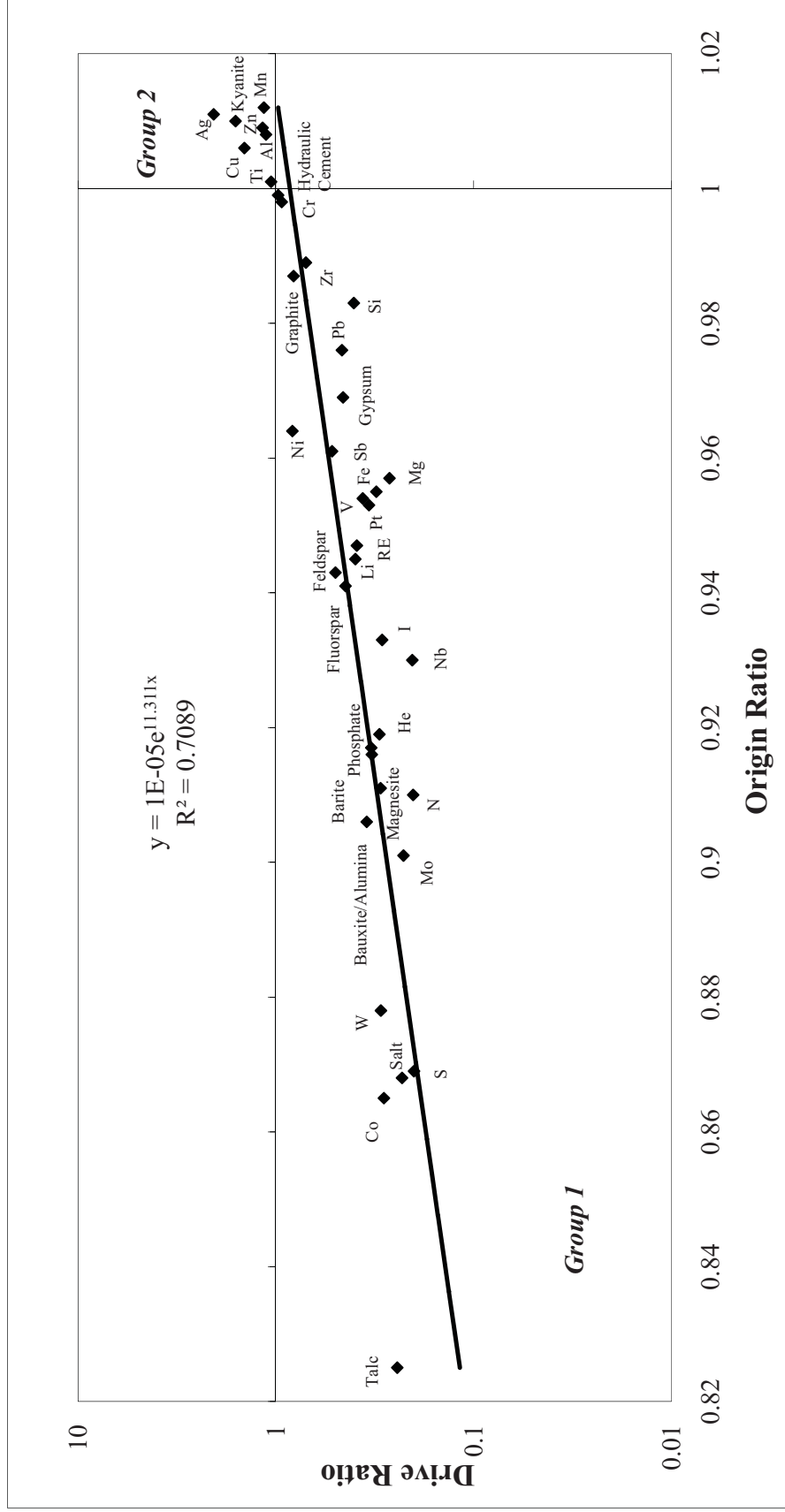


Figure 12(a). Engineering Materials Origin vs. Drive Ratio. Drive Ratio displaying relative strength of driving force of either patents or production activity. The driving force of nitrogen is strongest for a material whose production activity is driven by its patents since its ratio is farthest below one. The driving force of silver is strongest for a material whose patents are possibly driven by its production activity since its ratio is farthest above one. Alternately, the difference could be couched in terms of a hypothesis that elements like Niobium are driven by technical reasons or in other words, the innovation is highly technically influenced whereas the influence of new material technology on innovation diminishes for materials like Kyanite and Cu. Note also that the cross over point occurs at 1 (y-axis). The origin shift is the shift described in section 3 between the best-fit activity and best-fit patent evaluations for each material using the common pattern equation (1).

Fig. 12(a) appears to divide the materials evaluated into two groups. Group 1 is composed of materials whose patent activity is driving their production as suggested by the lag in production. The remaining materials in Group 2, are those in which patenting is driven by production suggested, conversely, by a lag in patenting. Stage IV materials do not fit into either of these groups and are possibly commodities in the full sense of the word whose pricing is fully set by demand and supply and with no supplier having a technological edge and no driving force associated with them.

Group 1 materials, according to their drive ratios had more than one patent published per normalized unit of production where patents may be thought to *drive* production activity. Nitrogen, for example, had one patent published per 0.202 normalized units of production. Group 2 materials had less than one patent published per normalized unit of production where the patents are possibly *driven* by production. For instance, silver had one patent published per 2.060 normalized units of production according to its drive ratio. In other words for each normalized unit of nitrogen production, 4.95 patents are found to be driving the production and for each normalized unit of silver production 0.49 patents are being driven by that unit of production. These results may be interpreted to mean that Group 1 is more innovatively active since more patents are required to drive one unit of production than Group 2 where production drives patents. Group 2 still is still innovatively active, but not as much as Group 1.

Alternatively, Fig. 12(b) presents materials whose production activity R^2 values are higher than 0.85 as opposed to Fig. 12(a) that includes all calculated values. The overall R^2 for the equation of the plot for 12(b) is higher than that of 12(a) indicating that higher best-fit R^2 values may predict a more accurate picture of what materials are in Stage III of their life cycles.

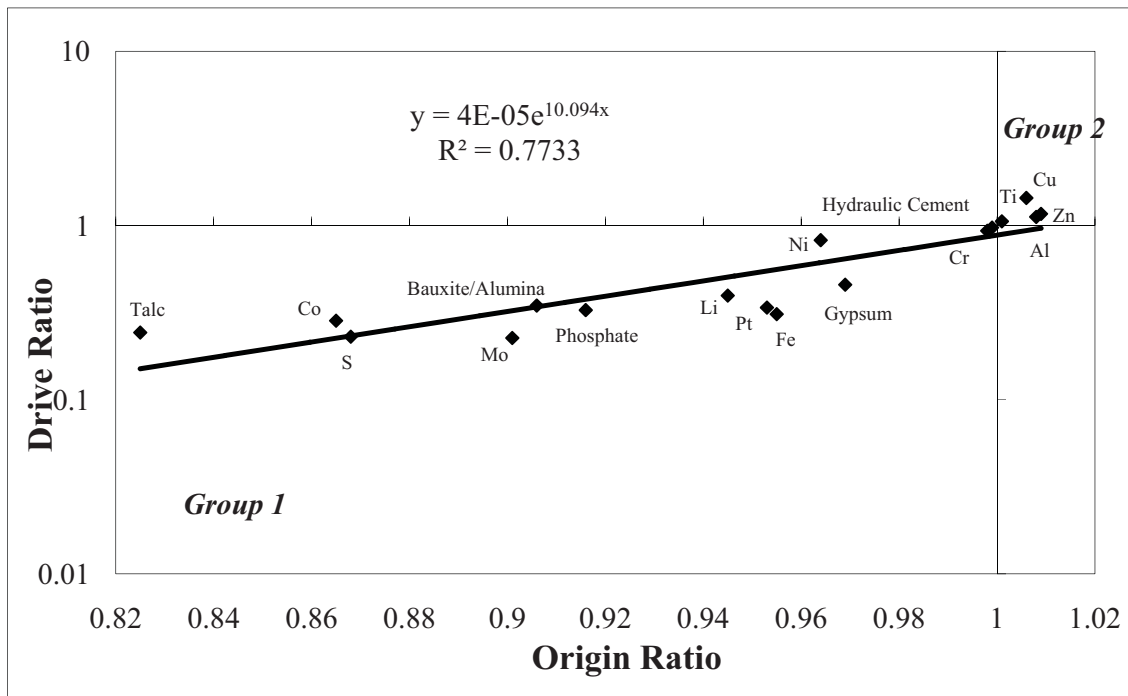


Figure 12(b). Engineering Materials With R^2 Over 0.85 Origin Ratio vs. Drive Ratio. Displays relative strength of driving force of either patents or production activity of materials with production activity R^2 values of over 0.85. This figure has higher overall R^2 value than Fig. 12(a) and crosses y-axis near one.

Figure 12(a) presents all materials whose data were capable of revealing best-fit R^2 values for their production data, then an R^2 value and an origin shift for their patent data when plugged into the same best-fit equation. An origin shift presents strong evidence of a Stage III material since a shift in origin displays driving or driven innovation and innovative activity is most strongly associated with Stage III. Better evidence of Stage III may be presented when a best-fit equation is run with the patent data first, resulting in a second patent R^2 value. Then the production data is plugged into the equation, leading to a second production R^2 and resulting origin shift. For the thirty-eight materials in Table 6, evaluation as just described led to twenty-one materials having new R^2 values for patents and production as well as a new origin shift in the opposite direction of the original as shown in Table 6. The

data and evaluations of the remaining seventeen materials would not support a second set of R^2 values.

Table 7. Origin Ratios, Shifts and R^2 . 1st production and patent R^2 represent the case where a production best-fit equation and R^2 were established then patent data was plugged into the equation resulting in a patent R^2 and an origin shift. Likewise, 2nd production and patent R^2 represent the case where a patent best-fit equation and R^2 were established first, rather than production as was done first previously, then production data was plugged into the equation resulting in a production R^2 and an origin shift. The 2nd shifts for all materials were in the opposite directions than the 1st shifts. The Avg. shift is the average of the absolute values of the two shifts, 1st and 2nd, with the sign of the 1st shift being employed in all cases for consistency. As a convention the sign chosen was the same as the 1st shift, but could be the sign of the 2nd if all are kept consistent.

Material	Production (1 st) R^2	Modified Patent (1 st) R^2	Shift (1 st)	Patent (2 nd) R^2	Modified Production (2 nd) R^2	Shift (2 nd)	Avg. Shift	Origin Ratio
Aluminum	0.9818	0.9658	(+)15	0.9554	0.9562	(-)17	(+)16	1.008
Antimony	0.7192	0.7644	(-)75	0.6640	0.9076	(+)65	(-)70	0.961
Bauxite	0.9521	0.4465	(-)179	0.9074	0.3453	(+)50	(-)114.5	0.906
Chromium	0.9731	0.9320	(+)3	0.9368	0.9698	(+)4	(-)3.5	0.998
Copper	0.9576	0.9320	(+)11	0.9416	0.9728	(-)13	(+)12	1.006
Feldspar	0.7833	0.4073	(-)108	0.7775	0.9789	(+)48	(-)78	0.943
Graphite	0.8122	0.9045	(-)24	0.9617	0.6803	(+)21	(-)22.5	0.987
Gypsum	0.8740	0.7123	(-)59	0.9543	0.6674	(+)42	(-)50.5	0.969
Hyd. Cem.	0.9743	0.9280	(-)1	0.9326	0.9732	(+)2	(-)1.5	0.999
Iron	0.8599	0.6888	(-)89	0.9329	0.0501	(+)49	(-)69	0.953
Kyanite	0.8387	0.8825	(+)20	0.9040	0.9293	(-)28	(+)24	1.010
Lead	0.5368	0.7303	(-)41	0.9172	0.1267	(+)40	(-)40.5	0.979
Manganese	0.5782	0.9666	(+)23	0.9132	0.7614	(-)36	(+)29.5	1.012
Nickel	0.9823	0.7331	(-)69	0.9502	0.7436	(+)41	(-)55	0.964
Platinum	0.9539	0.6065	(-)90	0.9121	0.9370	(+)48	(-)69	0.953
Rare Earths	0.8256	0.4483	(-)101	0.8152	0.8547	(+)43	(-)72	0.947
Silicon	0.6029	0.8980	(-)30	0.6195	0.6801	(+)31	(-)30.5	0.984
Silver	0.5027	0.3484	(+)21	0.7582	0.5563	(-)51	(+)36	1.011
Titanium	0.8600	0.9630	(+)1	0.9620	0.8774	(-)3	(+)2	1.001
Zinc	0.8805	0.9669	(+)18	0.9617	0.9650	(-)24	(+)21	1.009
Zirconium	0.6913	0.9536	(-)21	0.8964	0.4545	(+)17	(-)19	0.989

Table 7 contains two sets of two R^2 values and their resulting origin shifts and ratios as well as the average of the absolute values of the shifts. Origin shifts and R^2 values were sought by evaluating the production data first, then plugging in the patent data and then doing the same using the patent data first. In this way, four R^2 values are obtained as well as two origins, which have opposite signs. The opposite signs indicate that in one case the production or patents are the driving force and in the other case they are the driven force. For example, aluminum's 1st R^2 values indicate that its production is driving its patents by the

positive origin shift. The 2nd R^2 values, on the other hand, reveal that the patents are being driven by the production as shown by the negative shift in origin. Each set of R^2 values is revealing the same behavior, but from opposite directions. This leads to the averaging of the absolute value of the shifts, which may reveal a more accurate picture of the materials origin shift, innovative behavior and reaffirm the material as Stage III.

Figure 12(c) presents the shifts of the materials found in Table 7 and another method to express the drive ratio and evidence of Stage III. The results in it are similar to Fig. 12(a) and 12(b) with an exponential Drive Ratio/Origin Ratio plot that crosses the logarithmic y-axis near one and a fairly constant slope. The three Drive Ratio/Origin Ratio plots indicate materials that are in Stage III of their life cycles due to the presence of innovative driving forces coupled with the idea that innovative behavior is strongest in Stage III.

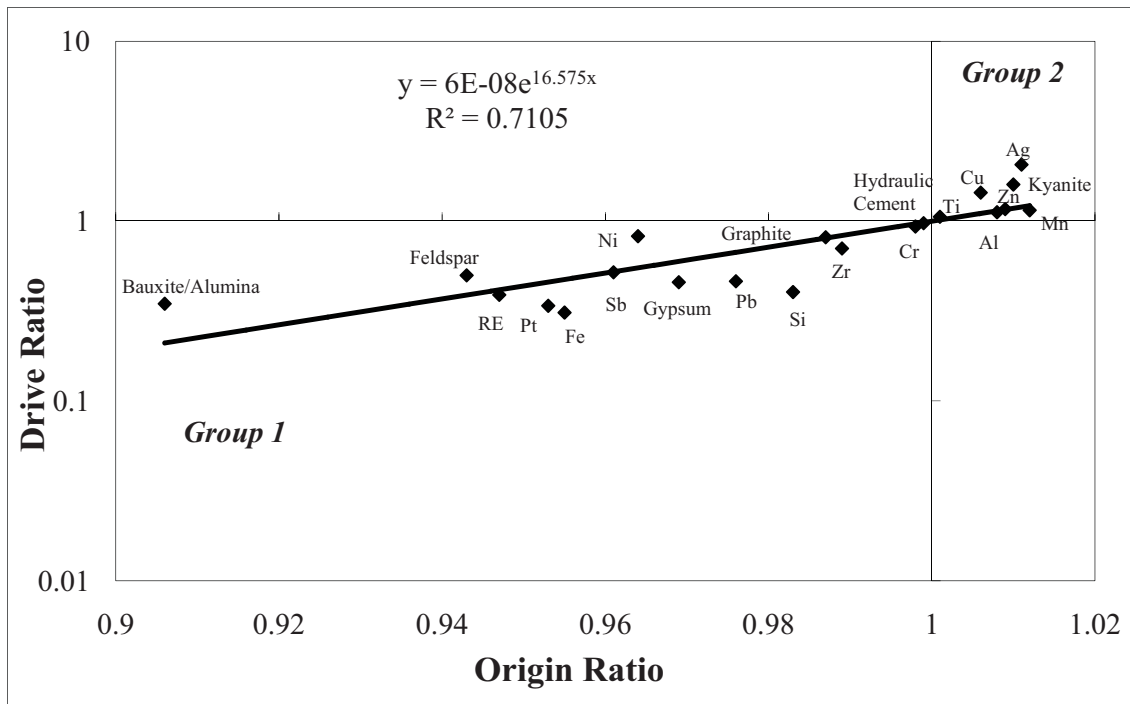


Figure 12(c). Engineering Materials With Two R^2 and Origin Shifts Origin Ratio vs. Drive Ratio. Displays relative strength of driving force of either patents or production activity of materials with two sets of two R^2 values and two origin shifts. Again, the plot crosses the y-axis near one and an overall R^2 value near that of Fig. 12(a).

Section 6: Energy Sources

An important goal of this dissertation is the determination of whether energy sources behave in a similar pattern as do engineering materials and if they can be evaluated in a like manner. Due to climate change and scarcity of energy resources there is a great interest in the development, diffusion and innovation of renewable, sustainable and carbon friendly forms of energy as well as policies that may be adopted by governments to utilize these clean energy sources [92-126]. The application of life cycle, best-fit and origin shift analyses as well as correlation theory to such energy sources would be a valuable addition to the understanding of their behavior. Accordingly, these procedures were applied to the energy sources listed in Table 2. Following the application of correlation theory, best-fit and origin shift analyses to engineering materials the same procedures were performed on the energy sources in Table 2. Due to the availability of consistent and continuous data sets, United States production for energy sources was chosen. Worldwide statistics were not available for the years to be researched on a complete basis and for many of the systems investigated there was no data.

Data Collection

The patent data for energy sources and materials was collected from the European Patent Office (EPO) web site in the same manner as described previously for engineering materials. Worldwide searches of patent counts per year were performed for each energy source using keywords in the titles and abstracts of published patents. Worldwide patent counts were made, rather than patent counts from only the USPTO, due to the existence of patent treaties which result in the effects of the innovation, that are represented by patents, being more global in scope. The global scope would lead to U.S. innovation relating to the U.S. production of these energy sources being affected by the honoring of patents from other

PCT member nations making worldwide patent counts legitimate for the purposes of this dissertation.

Production activity data for these systems and materials were collected from the U.S. Energy Information Administration (EIA), which is affiliated with the United States Department of Energy [91]. The data found on this site from this agency is consistent and complete for the years and energy sources required. The energy source production data for all energy sources is reported in billion or quadrillion Btu and then converted to kilo joules (kJ). Detailed production definitions can be found in Appendix 4.

Patent and Production Activity Data Correlation

The selected energy sources were subjected to the same correlation evaluations that were performed above on engineering materials. Correlation was sought between data sets relating to the production of kJ for energy sources, per year, and data sets composed of patents published per year for the same energy source. As with the evaluated engineering materials, a coefficient constant, r , and $100r^2$ were generated for each energy source listed in Table 8 below. Comparative plots of the generated production and activity were generated as well and can illustrate graphically the extent of any correlation that is present. Examples of such graphical illustration are found in Figs. 13 and 14. Figure 13 presents the correlation between the production activity and patents of wind energy and reveals a strong correlation and coefficient, r , of 0.9681. Figure 14 represents the correlation of the production activity and patent data of U.S. oil energy production and indicates a lack of correlation or coefficient, r , of -0.4862.

Table 8 reveals that there exists some degree of correlation for the production and patent data sets for energy sources except for U.S. oil energy. A relationship is thus exhibited between the production and patenting or innovation of energy resources allowing an

assumption that changes in one data set cause or are caused by changes in the other set of data as was shown for engineering materials. Further analysis of this data is thereby more reliable due to this proven correlative relationship. It is also apparent that energy sources behave in a similar manner as engineering materials when correlation theory is applied.

Table 8. Energy Sources Correlation Coefficients r and $100r^2$. Some correlation between production and patents exists to some degree for all energy sources except U.S. oil energy.

Material	Overall Correlation Coefficient r	$(100)r^2$
US Biofuel Energy	0.9469	89.67%
US Biomass Energy	0.5624	31.63%
US Coal Energy	0.8235	67.82%
US Fossil Fuel Energy	0.5956	35.50%
US Geothermal Energy	0.6660	44.36%
US Hydroelectric Energy	0.3617	13.0%
US Natural Gas Energy	0.4728	22.36%
US Nuclear Energy	0.8541	72.95%
US Oil Energy	-0.4862	-23.64%
US Renewable Energy	0.4030	16.24%
US Solar Energy	0.4786	27.90%
US Total Energy	0.8064	65.02%
US Wind Energy	0.9681	93.72%
US Wood Energy	0.8271	68.40%

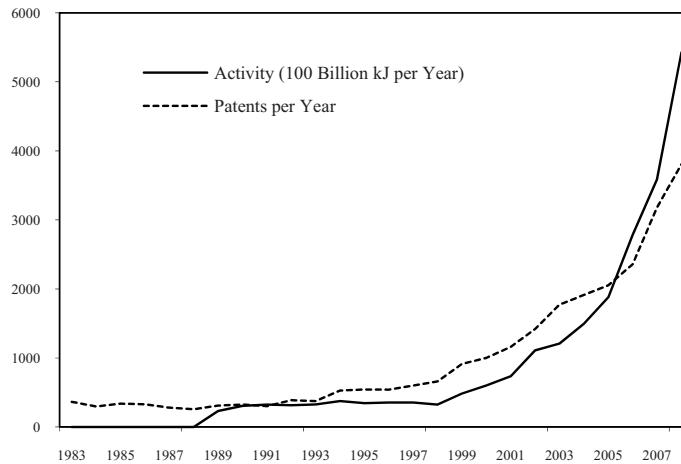


Figure 13. U.S. Wind Energy Activity and Patents. A strong correlation is illustrated by this figure. The calculated correlation coefficient was 0.9681. Data scaled to fit on the same figure.

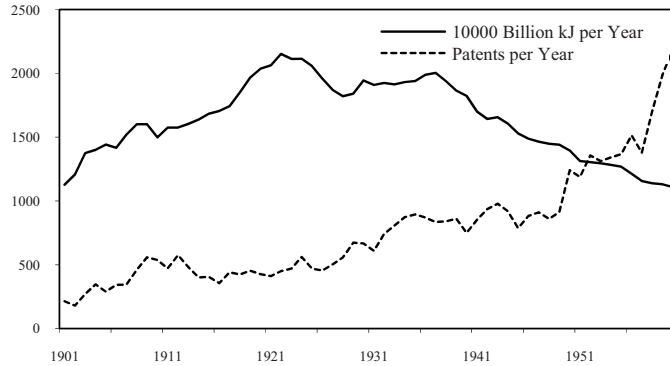


Figure 14. U.S. Oil Energy Activity and Patents. No correlation is exhibited in this figure with the two curves showing little resemblance. The calculated correlation coefficient was -0.4862. Data scaled to fit on the same figure.

Best-Fit

The best-fit common pattern equation (1) was applied, using a MatLab program, to energy sources data in an identical manner as was employed with engineering materials. As was the case with engineering materials, energy sources produced similar four-stage life cycles for production and patent data sets. Figure 15 displays a typical long-term four-stage life cycle for the production activity of U.S. hydroelectric energy. Stages I-IV are all present as well as a Stage II hump and “valley of death.”

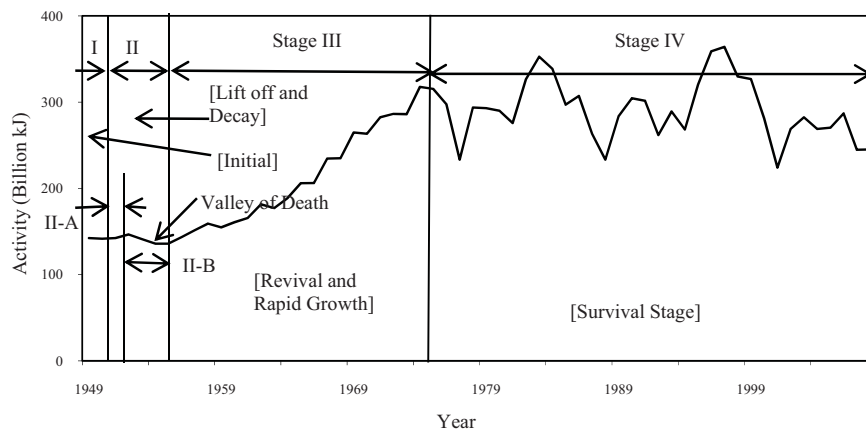


Figure 15. U.S Hydroelectric Energy. Illustration of a typical long-term life cycle for an energy source. The plot indicates the division of the life cycle into four stages that is common to engineering materials and specifically here to U.S. hydroelectric energy production.

Wind Energy as an Example. The common pattern equation and best-fit analysis were applied to fourteen energy sources as described in Section 4 for engineering materials. Wind energy serves as a good example of best-fit applied to energy sources. Figure 16 represents best-fit analysis for the production data of U.S. wind energy in thousands of kJ. The curve of the actual activity reveals that wind energy is likely in Stage III of its life cycle. Stages I-II are present along with a probable Stage III and rapid growth. There is no sign of leveling off or onset of the survival mode of Stage IV. A good R^2 of 0.8516 is present indicating a good fit. Figure 17 is composed of the best-fit analysis for the patent data of wind power. The life cycle for the patent data is also appears to be in Stage III as depicted by the strong rapid growth phase existing to the present. The patent data also has a good R^2 of 0.8183. All of the parameters in Figs. 16 and 17 are the same except for the origin, which has shifted negatively to 1961 for the patent data from the 1983 origin of the production data.

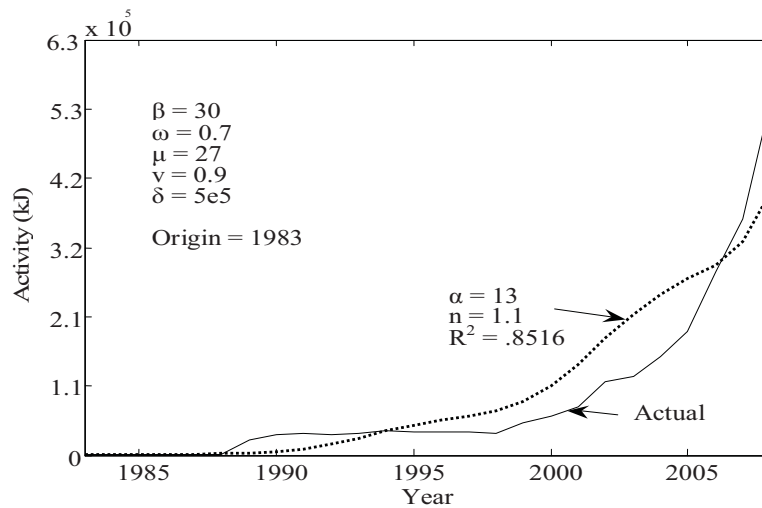


Figure 16. EIA U.S. Wind Energy Production. Fitted wind energy production activity curve with best-fit parameters and R^2 value and origin. The data is from EIA U.S. wind power production in kJ per year [91]. Shown in the figure are both the actual data curve and a best-fit curve.

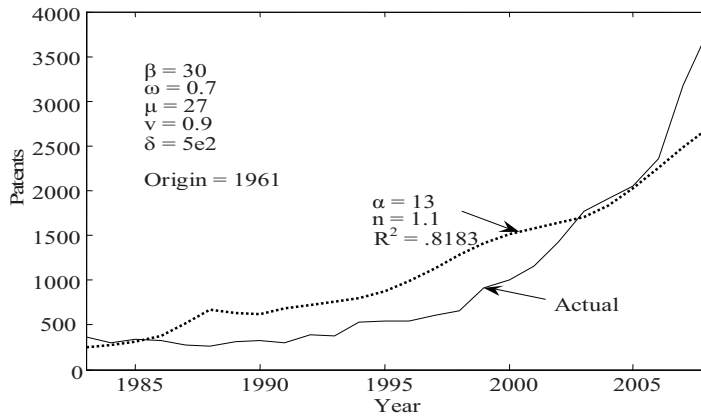


Figure 17. EPO Worldwide Patent Search: Wind and (Power or Energy) in Title or Abstract by Date of Publication. Wind energy modified patent curve with best-fit parameters and R^2 value and origin. The data is from the European Patent Office worldwide patents containing wind and (energy or power) in the title or abstract of the patent by date of publication. Parameters are the same as for the wind power production activity curve with only a shift in the origin.

The origin of 1961 for the wind energy patent best-fit curve signifies a 22-year lag for the production activity data when compared to patenting data. Twenty-two years after the patent data met a point in its life cycle, the activity data crossed the same normalized position in its own life cycle, which suggests that the patenting or innovative activity associated with wind energy is driving the production of the wind energy.

Table 9 below contains the individual origins and the origin shifts between the patent and activity curves for each of the energy sources evaluated. Wind, fossil fuel, solar, renewable, total, natural gas and coal energy displayed negative shifts of origin of 22, 38, 93, 97, 331, 392 and 428 years respectively. Such negative shifts indicate that the patenting curve reached a point in its life cycle a number of years before the production curve reached the same point in its life cycle. This could indicate that patenting or innovative behavior is *driving* production and displaying the constructive side of the innovative process. Figure 18 shows the negative 22 year origin shift for U.S. wind energy production. For wind power, a specific point that was reached by its activity production life cycle in 1990 was reached by its

patenting plot in 1978. The innovative behavior, represented by patents, thus occurred before the production of the wind energy. It could be then said that the innovation is causing or *driving* the production and that the innovation is strong and constructive which may be the case in a new technology where new processes and innovation are creating a need for new production.

Likewise, Table 9 shows that biofuel, biomass, geothermal and nuclear energy have positive shifts of 7, 8, 20 and 36 years, respectively, indicating that the patent curve reached a point in its life cycle a certain number of years after the production curve of the system reached the same point in its life cycle. This could indicate that the production of the energy source happens before the patenting or innovation and is being *driven* by the production. Such a behavior can be seen as an example of the destructive side of the innovative process. Figure 19 is an example of a positive origin shift of 20 years in U.S. geothermal energy. For geothermal energy, a certain point reached in the life cycle of the production activity that was reached in 1982, was likewise reached by the patenting plot in 2002. The patenting occurred twenty years after the production and thus was caused by or is being *driven* by the production of the system. Less strong innovative activity and the negative aspect of the innovative process is displayed in a case such as this and may occur in an older technology where innovation is being employed to better utilize production.

As with engineering materials, there are a number of indicators that point towards the life cycle stage that the energy source is in. Strong correlation is the first indicator of possible Stage III (Table 3), which is illustrated by a comparative graphical plot of the production and patent data. At times the plots relating to correlation reveal curves with Stage III features.

Positive best-fit R^2 values for either or both the production activity and patent data of an energy source are indicators of a possible Stage III as well. Such R^2 values are produced for curves that have Stage III life cycle attributes as shown in Fig. 3. Due to the fact that the

span of years for most energy sources is comparatively short when compared to engineering materials and the common pattern equation does not function optimally, lower values of R^2 are acceptable here where they would not for engineering materials.

The lower values of R^2 attainable for energy sources may be explained by the fact that the pattern equation is attempting to create a best-fit for a material or source with Stage III features. Sources with Stage I and II features may still generate an R^2 albeit a lower one. Stage IV sources need parabolic alphas and would not generate an R^2 with a linear alpha as have been employed here. Possibly, at the least, R^2 values generated for energy materials may suggest sources that are not in Stage IV but may be in Stage I, II or III.

Table 9. Energy Source Production and Patent R^2 values, correlation coefficient (r), origins, origin shifts and Stages. Energy sources are listed in order of descending Production R^2 values. As with engineering materials, an origin shift indicates the presence of a system in Stage III of its life cycle. Strong correlation and R^2 values near one are indicators of possible overall Stage III, but are not definitive evidence. Negative Production R^2 values, no correlation and lack of origin shift are indicative of Stage IV production energy sources. (* Indicates sources using Equation (2) for same scaling reasons as when used with engineering materials in Section 4.)

Energy Source	Production R^2	Independent Patent R^2	Modified Patent R^2	Correlation r	Production Origin	Patent Origin	Origin Shift	Stage
US Biofuel Energy	0.9024	0.8037	0.7881	0.9469	1981	1988	(+)7 years	III
US Coal Energy*	0.8547	0.5431	0.1697	0.8235	1949	1521	(-)428 years	III
US Wind Energy	0.8516	0.9097	0.8183	0.9681	1983	1961	(-)22 years	III
US Renewable Energy*	0.7404	0.3748	0.1237	0.4030	1949	1852	(-)97 years	III
US Biomass Energy*	0.7154	0.6243	0.6839	0.5624	1949	1957	(+)8 years	III
US Nuclear Energy	0.7142	0.1540	0.1758	0.8541	1957	1993	(+)36 years	III
US Geothermal Energy	0.6273	0.7138	0.4764	0.660	1960	1980	(+)20 years	III
US Total Energy*	0.3703	0.6511	0.2801	0.8064	1949	1618	(-)331 years	III
US Fossil Fuel Energy*	0.1995	0.7868	0.6993	0.5956	1949	1911	(-)38 years	III
US Solar Energy	0.1186	0.6355	0.4413	0.4786	1984	1891	(-)93 years	III
US Natural Gas Energy*	0.0989	0.8949	0.2094	0.4728	1949	1557	(-)392 years	III
US Wood Energy*	0.3728	0.9312	-	0.8271	1949	-	No Shift	IV
US Hydroelectric Energy*	0.1357	0.8470	-	0.3617	1949	-	No Shift	IV
US Oil Energy	Negative	0.5041	-	-0.4862	1949	-	No Shift	IV

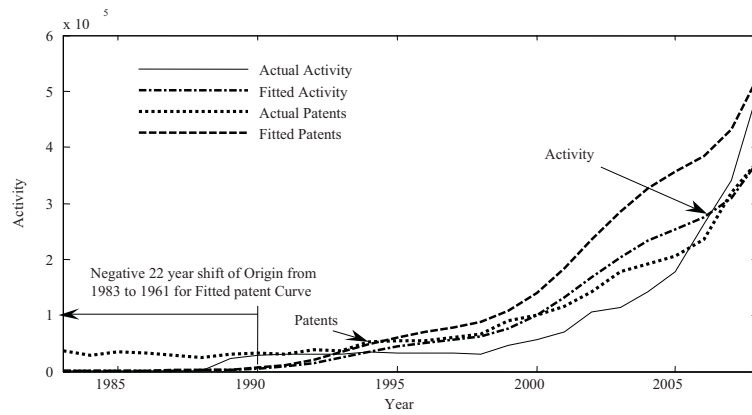


Figure 18. U.S. Wind Energy Best-Fit Activity and Patents. Plot showing the origin shift of patent and activity best-fit curves for wind energy. The shift is negative, indicating patent activity occurring before production activity and thus possibly driving the production. All parameters for the pattern equation are identical for the patent and production activity curves except for the difference in the origin that result in the negative origin shift.

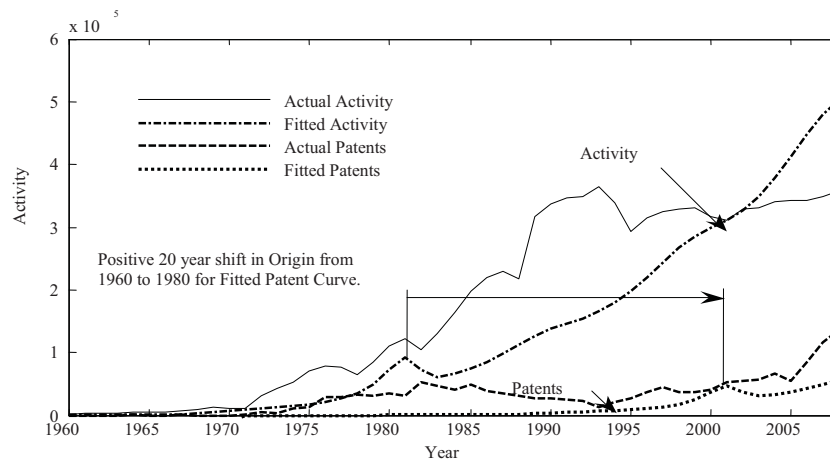


Figure 19. U.S. Geothermal Energy Best-Fit Activity and Patents. Plot depicting the origin shift of patent and activity best-fit curves for geothermal energy. The shift is positive, indicating patent activity occurring after production activity and thus possibly being driven by the production. All parameters for the pattern equation are identical for the patent and production activity curves except for the origins (the matching results in the positive origin shift).

Best-Fit, Origin Shift and Innovation

As with engineering materials the best-fit approach can be applied to comparisons of energy sources activity and patent data with patent and origin shifts, allowing inferences to be

made concerning the relationship of innovations to production for these energy sources. The best-fit equation and program were applied to all of the studied energy sources' activity and patent data independently. This analysis was performed successfully on the energy sources listed below in Table 10. Table 10 presents the α and n values for the independent activity and patent data and the original origin shifts and resulting origin ratios, $(x_0 + OS)/x_0$, derived for Stage III energy sources as well as α^n and the drive ratios, again defined as $(\alpha^n)_{activity} / (\alpha^n)_{patent}$. Similar to the drive ratio for engineering materials, the drive ratio generally becomes progressively larger than one, as the origin ratio becomes larger than one representing the origin shift moving away from zero in a positive shift direction. The drive ratio also approaches zero as the origin ratio becomes progressively smaller than one which is the result of the origin shift moving further in the negative direction from zero as shown in Fig. 20 below.

Table 10. Energy source α and n parameters, drive ratio and origin shifts, origin ratios and modified R^2 . Alpha and n are from the pattern equation. Ratio of α^n indicates strength of the driving force of the material. The farther the ratio is from one, in either direction, the greater the driving force. A positive origin shift could indicate patents being driven by production. A negative origin shift suggests production being driven by patents. Note that (n_d/n_p) is less than one for production sources with negative shifts in origin and one or above for sources with positive origin shifts. Modified R^2 generally becomes smaller than one as the origin ratio moves farther from one.

	α_a	n_a	α_a^n	α_p	n_p	α_p^n	Drive Ratio	n_d/n_p	Origin Shift	Origin Ratio	Modified R^2
Nuclear	13	1.3	28.06	16	1	16	1.75	1.3	+36	1.018	0.1758
Geothermal	17	0.8	9.65	31	0.5	9.65	1.73	1.6	+20	1.010	0.4764
Biomass	12	1	12	14	0.9	10.75	1.12	1.11	+8	1.004	0.6839
Biofuel	15	1.2	25.78	22	0.9	16.15	1.60	1.33	+7	1.004	0.7881
Wind	13	1.1	16.80	13	1.5	46.87	0.36	0.73	-22	0.989	0.8183
Fossil Fuel	36	0.29	2.83	41	1	41	0.069	0.29	-38	0.981	0.6993
Solar	37	0.7	12.52	39	1.4	168.9	0.074	0.5	-93	0.953	0.4413
Renewables	1	0.09	1	14	0.5	3.74	0.267	0.18	-97	0.950	0.1237
Total Energy	46	0.34	3.68	12	1.2	19.73	0.186	0.283	-331	0.830	0.2801
Nat. Gas	22	0.2	1.86	32	0.9	22.63	0.082	0.22	-392	0.799	0.2094
Coal	28	0.2	1.95	17	1	17	0.115	0.2	-428	0.780	0.1697

As with engineering materials the use of the pattern equation in the case of energy sources creates a relationship between α and n that can be evaluated and compared to origin

shifts produced by independent patent and production activity best-fit derivations. A graphical representation, such as Fig. 20, of the relative scale, or distance, of the origin shift can be made, indicating an absolute amount that the patent or activity driving force has on the other. Such a curve may effectively represent innovation behavior. Noted is that the activity of fossil fuel power is possibly being driven the most by its patents since its drive ratio is nearest to zero for systems whose activity is driven by patents. Similarly, the patents of nuclear power may be driven the most by its activity because its ratio is farthest from one for energy sources whose patents are driven by activity

Again, as with engineering materials three patterns emerge from the best-fit analysis for energy sources shown in Table 10. The first is that when a positive origin shift is indicated, the drive ratio is always above one (the ratio is always less than one and approaches zero when the origin shift is negative). Second, when n_a is divided by n_p the resulting ratio is always less than one for energy sources that have negative shifts in origin and the resulting ratio is always one or above for energy sources with a positive origin shift. Lastly, as the origin ratios move away from one in either direction the modified patent R^2 , which is generated by patent data being run with the common pattern equation production parameters, generally becomes smaller than one. As indicated in Table 10 and on Fig. 20 energy sources such as coal and natural gas energy have origin ratios the farthest below one and lower modified R^2 than do sources with ratios nearer to one. In the same way nuclear energy has the highest origin ratio and has an R^2 less than sources with origin ratios near to one. This could indicate that energy sources may approach Stage IV when they reach origin ratio extremes and enter stage IV when modified R^2 values become too small to support an origin shift or origin ratio in a manner similar to engineering materials.

Fig. 20 appears to divide the energy sources evaluated into two groups. Group 1, containing coal, natural gas, wind, renewable, fossil fuel, solar and total energy, is composed

of energy sources whose patent activity is driving their production as suggested by the lag in production. Biomass, biofuel, geothermal and nuclear energy are in Group 2, in which patenting is driven by production suggested, conversely, by a lag in patenting.

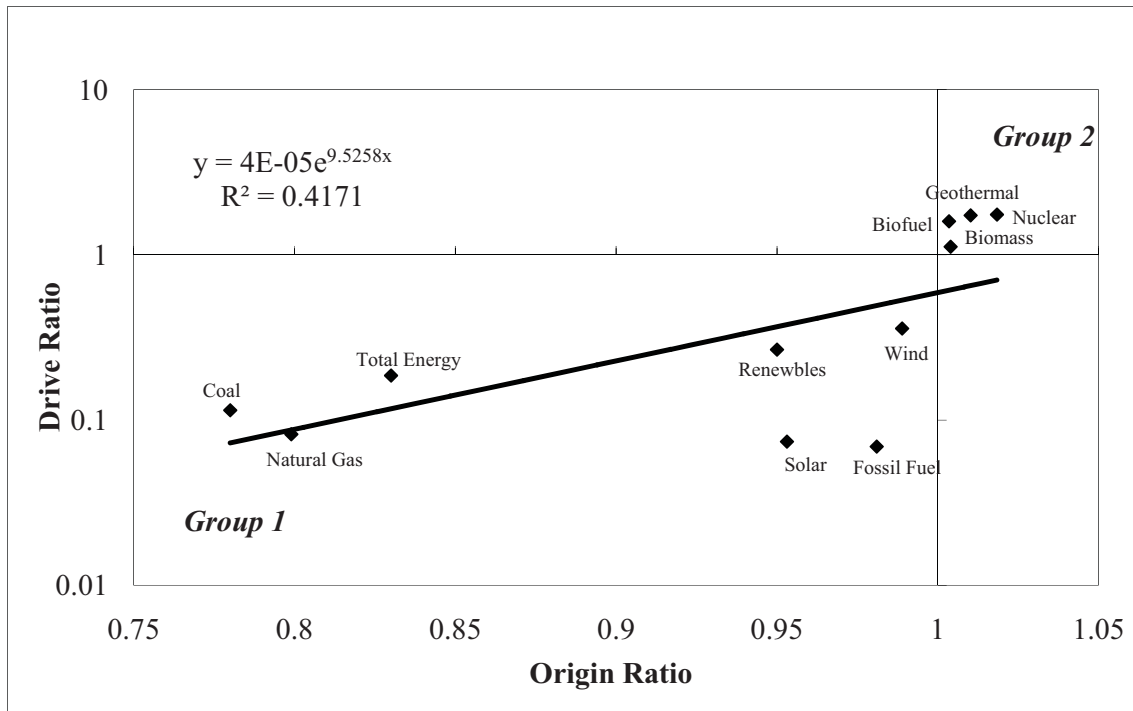


Figure 20. Energy Source Origin Ratio vs. Drive Origin. Displays relative strength of driving force of either patents or production activity.

Group 1 energy sources, according to their drive ratios had more than one patent published per normalized unit of production where patents may be thought to *drive* production activity. U.S. fossil fuel energy, for example, had one patent published per 0.069 normalized units of production. Group 2 energy sources had less than one patent published per normalized unit of production where the patents are possibly *driven* by production. U.S. nuclear energy had one patent published per 1.75 normalized units of production according to its drive ratio. In other words for each normalized unit of fossil fuel energy production, 14.5 patents were found to drive the production and for each normalized unit of nuclear energy

production 0.57 patents are being driven. These results may be interpreted to mean that Group 1 has more innovation associated with it since more patents are required to drive one unit of production than Group 2 where production drives patents. Group 2 is still innovatively active, but not as much as Group 1. And, like engineering materials, the presence of an origin shift in energy sources indicates constructive or destructive innovation, which is a classic feature of Stage III and thereby strongly points towards the presence of a Stage III source or at least an energy source that is not in Stage IV. It is possible that Stage I or II is in evidence for materials such as solar and wind energy that have had shorter life cycles and produce low R^2 values. These energy sources may be in the opening stages of their long-term life cycles. Their origin shifts indicate the presence of constructive or destructive innovation which could exist in Stages I and II as well as they do in Stage III, though to a lesser degree. In such cases an origin shift is good if not definite evidence of the presence of Stage III, but it is strong evidence of the absence of Stage IV.

Energy Materials

This section contains correlation, best-fit and origin shift analysis data for the production and patents of energy producing materials, in tons, barrels or cubic feet, of coal, natural gas, oil and uranium rather than energy (kJ) produced by them, which was done for energy sources above. The production data is from the EIA web site [91] and the patent data is from the EPO site [80]. All data gathering techniques, correlation, best-fit and origin shift analyses were carried out in an identical manner as for engineering materials and energy sources. Energy materials behaved in a similar manner as engineering and materials. Details concerning the gathered data and evaluations for energy materials can be found in Appendix 5. These four materials all exhibit Stage III behavior with origin shifts.

Table 11. Production and Patent R^2 values, correlation coefficients (r), origins, origin shifts and Stage for each evaluated energy materials. Equation 2 was used in all cases.

Energy Source	Production R^2	Independent Patent R^2	Modified Patent R^2	Correlation r	Production Origin	Patent Origin	Origin Shift	Stage
US Coal	0.7142	0.7523	0.6276	0.8368	1900	1870	(-)30 years	III
US Natural Gas	0.3701	0.9332	0.2536	0.6982	1936	1610	(-)326 years	III
US Oil	0.2334	0.9243	0.4524	0.5402	1900	1920	(+)20 years	III
US Uranium	0.1256	0.0159	.0154	0.6265	1949	1819	(-)130 years	III

Table 12. Energy material α and n parameters, drive ratio and origin shifts and ratios for energy materials. Materials arrange by descending modified R^2 .

	α_a	n_a	α_a^n	α_p	n_p	α_p^n	Drive Ratio	n_a/n_p	Origin Shift	Origin Ratio	Modified R^2
Coal	15	0.6	5.08	23	0.7	8.98	0.566	0.86	-30	0.984	0.6276
Oil	24	0.7	9.25	33	0.6	8.14	1.14	1.17	+20	1.011	0.4524
Nat. Gas	17	0.23	1.92	15	0.8	8.73	0.220	0.288	-326	0.832	0.2536
Uranium	2	0.29	1.22	10	0.3	1	0.61	0.97	-130	0.94	.0154

Table 11 reveals that oil has a positive origin shift and therefore in the destructive mode of innovation while coal, natural gas and uranium have negative shifts and are in the creative mode of the innovation process. Table 12 continues the trends displayed by engineering materials and energy sources of positive origin shifts producing origin ratios and drive ratios over one and negative shifts producing ratios below one. The modified R^2 also tends to be farther from one the larger the shift in origin is.

Figure 21 shows that oil is in group 2 and natural gas, coal and uranium are in group 1. By comparison in Fig. 20, nuclear energy is in group 2 while coal and natural gas energy are in Group 1 and oil energy is in Stage IV with no origin shift. No pattern emerges since only two of the materials have the same direction of shift as its energy counterpart. Similar behavior is exhibited, though, for energy systems as has been revealed for engineering materials and energy sources.

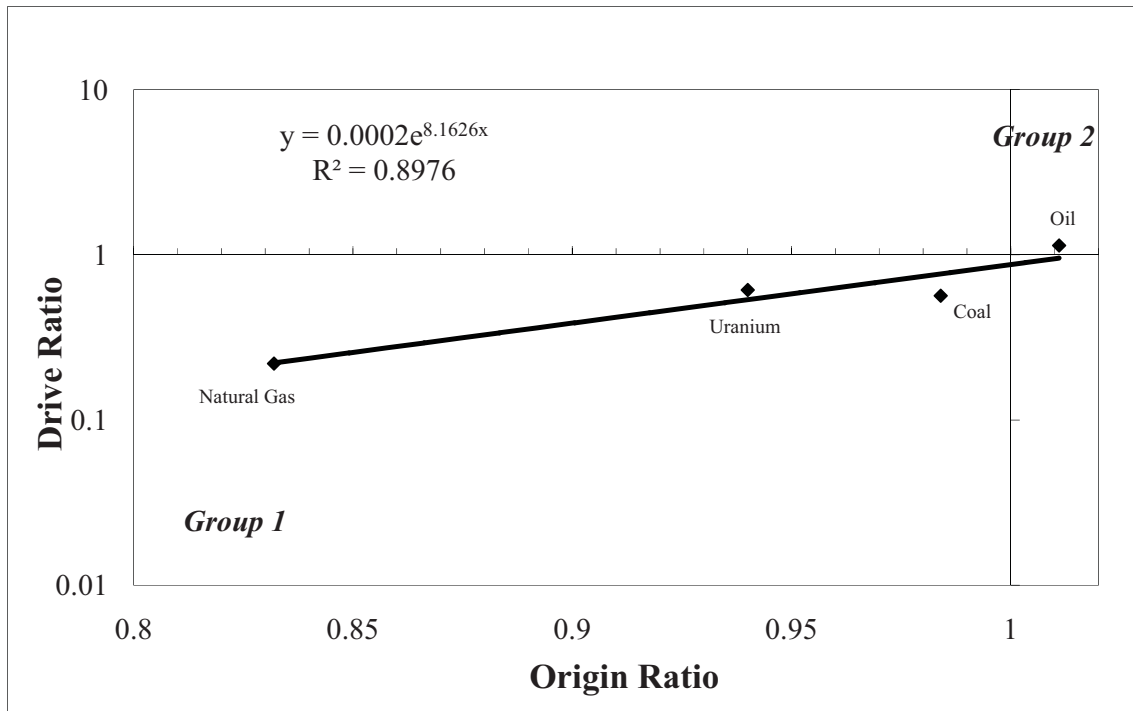


Figure 21. Energy Materials Origin Ratio vs. Drive Ratio. Displays relative strength of driving force of either patents or production activity for energy materials. R^2 is high probably due to only four points being in the plot.

Engineering Material Connection

In several cases the best-fit and correlation evaluations of energy sources revealed a connection between such energy sources that are predicted to be in Stage III, or innovatively active and engineering materials that are also innovatively active, or in Stage III, of their life cycles. Table 13 give examples of energy sources and the engineering materials that are both being innovated and helping to innovate the energy source. Vanadium and silicon are Stage III materials in the constructive mode and are commonly used in products related to solar and wind power production [90]. Other materials such as graphite, nickel, cobalt, silver, manganese, the rare earths and lithium, which are in Stage III, are being innovated largely in support of energy production and storage. Even Stage IV materials (cadmium, selenium, fluorspar) are being innovated in the energy generation field [90]. In three cases innovatively

active energy sources, coal energy, nuclear energy and natural gas energy use for fuel, coal, uranium and natural gas that are also in Stage III. With energy resources declining worldwide and environmental concerns, it seems logical that materials are being innovatively employed to meet energy requirements in an environmentally friendly manner resulting in Stage III growth and Stage IV stability for these materials. It appears that where the need for innovative energy sources arises the innovative behavior of engineering materials that might fill that need rise as well.

Table 13. Examples of energy sources and the engineering materials that are innovatively active possibly due to their usage in the related energy source [90].

Energy source	Related Material	Usage
<i>Solar Energy</i>	Vanadium	Vanadium Redox Batteries (large power storage)
"	Silicon, Selenium	Solar Cells
<i>Wind Energy</i>	Vanadium	Vanadium Redox Batteries (large power storage)
<i>Nuclear Energy</i>	Fluorspar	Nuclear Fuel Additive
"	Uranium	Fuel
<i>Renewable Energy</i>	Graphite	Fuel Cells, Batteries
"	Nickel, Rare Earths, Cobalt	Rechargeable Batteries
"	Lithium, Cadmium, Lead	Batteries
"	Manganese	Dry Cell Batteries
"	Silver	Battery Electrodes
<i>Coal Energy</i>	Coal	Fuel
<i>Natural Gas Energy</i>	Natural Gas	Fuel

Table 14. Comparison of origin shifts, origin ratios and drive ratios of energy sources and the engineering and energy materials that are related to them [90].

	Origin Shift	Origin Ratio	Drive Ratio
Solar Energy	-93 years	0.953	0.074
Vanadium	-90 years	0.954	0.363
Silicon	-30 years	0.985	0.435
Wind Energy	-22 years	0.989	0.36
Vanadium	-90 years	0.954	0.363
Nuclear energy	+36 years	1.018	1.75
Uranium	-130 years	0.94	0.61
Fluorspar	-113 years	0.941	0.444
Coal Energy	-428 years	0.780	0.115
Coal	-30 years	0.984	0.57
Natural Gas Energy	-392 years	0.799	0.082
Natural Gas	-326 years	0.832	0.22
Renewable Energy	-97 years	0.950	0.267
Graphite	-24 years	0.987	0.813
Nickel	-69 years	0.964	0.824
Rare Earths	-101 years	0.947	0.389

Cobalt	-256 years	0.865	0.284
Lithium	-106 years	0.945	0.396
Lead	-41 years	0.978	0.434
Manganese	+23 years	1.012	1.418
Silver	+21 years	1.011	2.060

Table 14 indicates that in a majority of cases the origin shift direction of the materials follows that of their particular related energy source. Likewise, the origin and drive ratios are higher or lower than one for the material as that of its related energy source. Nuclear energy is the only obvious exception to this pattern, however it has only two materials to compare it with. The degree of the shift or ratio of the materials seem to have little relation to that of the energy sources. It still can be said that innovatively active, or Stage III, energy sources utilize or are partially enabled by engineering and energy materials that are also innovatively active

Energy Source and Engineering Material Result Comparison

The application of correlation theory and best-fit analysis to energy sources as was applied to engineering materials has disclosed that energy sources behave in a similar manner to engineering materials. In both cases, correlation theory has revealed that production and patent data have a relationship to each other and that changes in one set of data have an impact on changes in the other data set. Energy sources have been shown to display the same four-stage life cycle exhibited by engineering materials in both production and patent data. Origin shift and innovative patterns are found in each case as well. Positive shifts are always found where the drive ratio is greater than one and negative when the drive ratio is less than one. Also, the ratio produced by dividing n_a by n_p is less than one where a negative origin shift is found and one or above where a positive origin shift exists.

The drive ratio versus origin shift plots (Figs. 12 (a-c), 20 and 21) for energy sources, engineering materials and energy materials all reveal a possible connection to the dual nature of innovation as proposed by Schumpeter [55]. Engineering materials and energy sources

with drive ratios below one indicate production drive by innovation and the constructive nature of the innovative process. Energy sources and engineering materials with drive ratios above one represent the destructive facet of the innovative process where production drives innovation as measured by patents. Innovation is present in both instances but is stronger on the constructive side. Finally, Stage III is indicated in both engineering materials and energy sources where a shift in origin is found through best-fit analysis.

Section 7: Analysis

Schumpeterian economic theory posits the idea that innovation propels capitalistic economies through a process of “Creative Destruction” where innovation constantly destroys the old while it creates anew [55]. Patents, being analogous to innovation, may be considered as tools to carry out this process of “Creative Destruction.” Patents can be used to prevent innovation or deny the right to compete when used to protect property rights by blocking the technical innovation of others in the same industry. This is the destructive side of patents. The creative side is illustrated by the use of innovation to overcome protective patents, which result in more patents and new innovative products. Numerical and graphical proof that patents can indicate and measure the destructive and creative functions that innovation can exhibit is offered here.

Correlation

Correlation, to some degree, was shown to exist between patent and activity data for forty-eight of fifty evaluated engineering materials and thirteen of fourteen energy sources. Thus, statistically, the number of patents published, in reference to an engineering material or energy source, is often correlated at least to some degree to the amount of production for that material or energy source on a yearly basis. Materials and energy sources with strong

correlation (Al, Cr, Ni, wind power, e.g.) often appeared to be in Stage III of their life cycles while those with weak or no correlation (Hg, Be, As, hydroelectric power, e.g.) are commonly considered to be in Stage IV. The correlation coefficient r gives evidence that variations in one of the data sets, activity or patent, can be attributed to variations in the other data set. In other words, change in one set of data drives the change in the other set. Comparisons of the production and patent data are thus more relevant due to the relationship that exists between them, which correlation theory indicates. Best-fit analysis can aid in determining which data came first and thus, drove the change in the other set.

Best-Fit

The best-fit equation and program can identify the four stages in the life of the production and patent data of an engineering material or an energy source. Most innovative activity occurs during Stage III and appears to diminish or cease at the onset of Stage IV, making the identification of these stages an important objective. The curves and coefficients produced via correlation analysis are a start in identifying the stage of a material or system. There is an obvious flattening of the curves during Stage IV when compared to Stage III. If the correlation curve flattens for an extended period and the correlation coefficient drops, the material or system is likely in Stage IV. Another indicator of the stage of a material or system is its R^2 value. The materials and systems studied here show that, in general, Stage III items have R^2 values for production and patenting data approaching one for engineering materials and positive R^2 values for energy sources and materials. The strongest and most obvious Stage III materials and systems have R^2 values approaching one generated from the production, modified patent and independent patent best-fit evaluations such as nickel, aluminum and wind power. Stage IV is very strongly indicated when R^2 is negative or near negative such as for cadmium, arsenic, beryllium and fossil fuel power whose generated life

cycle curves back up this inference. The combination of no correlation, a negative or near negative R^2 and life cycle curves that illustrate a steep sustained drop off of production, as for mercury, may indicate a Stage V or “Final Death” where innovative activity is very low again reaching the initial starting levels of its Stage 1.

Stage V

As suggested previously in section 2, a Stage V may exist in the life cycle of materials. Figure 22 below shows a proposed Stage V occurring after Stage IV for mercury. Stage V exhibits a steady decline in production with no clear sign from the data of revival as exists in Stages III or IV. With no clear sign of revival, Stage V can be called the “Final Death” stage. Such “final” death may be due solely to resource depletion. However, environmental and health concerns for toxic substances, such as asbestos, mercury and beryllium could also explain a decrease in production [90]. A final explanation for death could be the replacement of a material with another less costly one or the successful innovating around a toxic or expensive material thereby causing a decrease in production. While the onset of Stage V may be caused by resource depletion, the materials investigated here that display Stage V behavior are asbestos, beryllium and mercury, which are toxic substances with environmental concerns attached to them. Arsenic and cadmium, also toxic materials, are now Stage IV but seem to be approaching Stage V. Such environmental concerns seem to be a more likely cause, rather than resource depletion, for the onset of Stage V according to the results of this study.

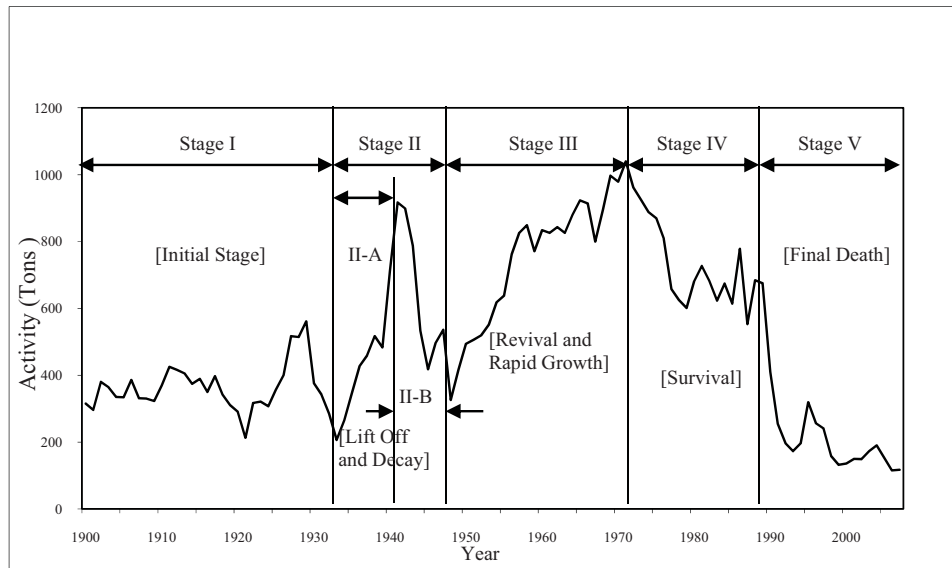


Figure 22. Mercury Production Life Cycle. The life cycle curve for mercury depicting it as a Stage V material. This figure identifies the four stages as discussed in Section 2 and shown in Fig. 3. Also shown is the possible Stage V or “Final Death Stage”. Here Stage V begins around 1988 and is identified by a steep decline in production with little clear sign from the production data for any future stabilization or revival as is often displayed in Stages III and IV. Stage V could indicate resource depletion or less demand for the material or innovating away from the material due to environmental or toxicity concerns such as exist with asbestos and mercury. Activity is in tens of tons.

Origin Shift

The best-fit analysis of the data can confirm and further identify the time frames where Stages III and IV may exist. The best-fit equation was applied to the patent data of each engineering material and energy source with the same pattern equation and parameters, as was previously done to the production data. A change or shift was sought in the data origin in these modified patent best-fit equations. Positive and negative shifts were each found for engineering materials and energy sources. A negative shift implies that the patenting occurs previous to the activity and therefore variations in the patents may drive change in the activity. A positive origin shift indicates that activity occurs first and variations in it possibly drive variations in the patents. Also, the presence of an origin shift indicates Stage III since a shift possibly predicts the mode or direction of innovative activity and such activity is

strongest in Stage III, which in these materials and systems is backed, for the most part, by Stage III type curves, good correlations and high R^2 values.

Driving Force

Driving force behavior can be illustrated by best-fit analysis and comparisons of α and n parameters with origin ratios of these thirty-eight materials and twelve energy sources. Nitrogen appears to be, amongst the engineering materials studied here, the one whose patents most drive its production activity. Silver, on the other hand, is the material whose activity drives its patenting the most. Nitrogen, along with the other group 1 materials (Fig. 12(a)), have a drive ratio less than the universal constant n_0 of one and could be thought to be in the creative mode of the innovative process where patents and innovation spur production and economic growth. Silver and the remaining group 2 materials are in the destructive or negative mode of the innovative process with drive ratios above the universal constant, n_0 , of one where patents could be used as protection of property rights and as a result stifle innovation and possibly economic growth. Materials with low drive ratios may be in the position of being farthest into the creative part of the innovative process due to their use in the electronics, metallurgical or other high-tech industries [90]. Patenting and innovation are very focused for materials in these areas. High drive ratio materials are used for a wide variety of applications in lower-tech industries [90]. The activity of such materials may be driving the patenting because the variety of products made from them are less cutting-edge and high-tech than those of a material, such as nickel, whose patents drive its production activity.

Examples of the uses of specific engineering materials seem to support the idea that negative origin shifts and high alpha ratios indicate more innovative activity while positive shifts indicate less. Aluminum, copper, manganese, titanium and zinc are used in a great

variety of applications and lack a narrow focus of innovation and patenting though they do have some high-tech applications [90]. Materials with negative origin shifts, or low origin ratios, and low alpha ratios often have application in very focused, high-tech and currently highly innovative areas. Nickel is used for super alloys, aerospace and rechargeable batteries [90]. Silicon is employed in solar cells and semiconductors, the rare earths are important in rechargeable batteries and electronics and iodine is a component of LCDs, which are very important in the electronics industry [90]. Though there are some anomalies, the general pattern illustrated is that that lower alpha ratios and lower origin ratios are signs of innovatively stronger materials while less innovative materials have higher ratios and shifts.

U.S. fossil fuel energy appears to be the energy source whose patents most drive its production while U.S. nuclear power is the energy source whose patents are most driven by its production. As with engineering materials, the group 1 energy sources of renewable, wind, coal, natural gas, fossil fuel, solar and total energy, have drive ratios less than the universal constant, n_0 , of one and are systems that appear to be relatively high-tech and in currently highly innovative areas [92-99, 106-110]. Group 2 energy sources, biofuel, biomass, nuclear and geothermal energy, have drive ratios above the universal constant of one also in the same manner of engineering materials and could be said to be in less high-tech areas [100-105]. Energy sources in group 1 could be said to be in the constructive phase of innovation since their production is being driven by innovation as measured by patents while those in group 2, where production drives innovation, are in the destructive segment of the innovative process.

Origin ratios below one represent negative origin shifts and the constructive mode of the innovative process. However, some engineering materials and energy sources seem not to have the cutting edge applications that would merit a high degree of innovation but are nonetheless in Group 1. Materials such as talc, salt and sulfur and the energy sources of coal and total energy have some of the lowest origin ratios of those evaluated. These five materials

and energy sources and others with low origin ratios also, in general, also have lower modified patent R^2 values as shown in Figs. 6 and 10 than do those with origin ratios near one. Such relatively poorer fits of patent data to production activity best-fit parameters may indicate that the resulting innovative activity is mainly to keep the production afloat instead of developing new innovative uses of or improvements to the material or energy source. A material or energy source may enter Stage IV when the modified patent R^2 becomes so low that it cannot generate a shift in origin and thereby an origin ratio. Such a point could illustrate where innovation can no longer keep the production afloat and Stage IV sets in.

Origin ratios above one indicate the destructive aspect of the innovative process where production is driving the patenting. Engineering materials and energy sources in Group 2 have drive ratios above one, origin shifts above one and modified patent R^2 values that decrease from one as the origin shift grows larger than one. Silver and nuclear energy are the material and energy source, respectively, that have the highest origin ratio and lowest R^2 value for group 2 materials and energy sources. Silver and nuclear energy may be nearest to having no innovation for production to drive and thus to Stage IV. This may be the case since as the R^2 values decrease in Group 2 they will not be able to produce an origin shift or ratio.

Analysis of the driving force may reveal two means that a material or energy source may leave Stage III and enter Stage IV. Stage IV seems to be approached when a material or energy sources nears either end of the trend lines of Figs. 12(a-c) and 20. In either case it appears that the driving force is losing the driven production or patenting. In Group 2 the materials or energy sources at the extreme of the trend lines have production with little innovation to drive, are not innovatively active, leave Stage III and becomes commodities. With Group 1, materials or sources at the extreme of the trend lines, have little production for the innovative activity to drive and may also enter Stage IV.

Life Cycle Stage Change

The results generated by the processes described here are not static for the tested materials and sources but instead are of a fluid nature. The r , R^2 , stage, origin shift or driving force of a material can change over time. A previous study of twelve metals, utilizing production and patent data for the years 1900-2004 produced results that in some cases were quite different from the results generated in this dissertation which covers the years 1900-2007 [127]. Table 15 indicates that of the twelve metals evaluated in the previous study, four of them (iron, manganese, molybdenum and tungsten), changed from being Stage IV to Stage III.

Table 15. Comparison of the current 1900-2007 results with previous results for twelve metals for the years 1900-2004 [92]. The stage in the life cycle of the material as well as its origin shift and r and R^2 values for each span of years are recorded.

Result Changes Between 1900-2004 and 1900-2007 Data								
Material	Estimated Stage		Correlation r		Production Best-Fit R^2		Origin Shift	
	2004	2007	2004	2007	2004	2007	2004	2007
Aluminum	III	III	.9623	.9652	.9801	.9818	+15	+15
Chromium	III	III	.9483	.9495	.9690	.9731	-2	-3
Copper	III	III	.9430	.9507	.9397	.9576	+11	+11
Iron	IV	III	.8682	.8741	.5162	.8599	-	-89
Magnesium	III	III	.8817	.9078	.6467	.7502	-79	-83
Manganese	IV	III	.6312	.6835	.6140	.5728	-	-23
Molybdenum	IV	III	.9184	.9289	.9430	.9538	-	-188
Nickel	III	III	.9525	.9563	.9676	.9823	-57	-69
Titanium	III	III	.9011	.9150	.8303	.8600	+2	+1
Tungsten	IV	III	.7587	.7419	.5528	.6449	-	-232
Zinc	III	III	.9249	.9387	.8655	.8805	+26	+18
Zirconium	III	III	.8900	.9239	.5520	.6913	-22	-21

Manganese production for the years 1900-2004 is displayed in Fig. 23. In 2004 Mn was in Stage IV with its production oscillating after peak production in 1976. 2004 production had not reached the peak production of 1976. Production was rising and falling during these Stage IV years of 1976-2004 with no clear signs of a continual rise in production.

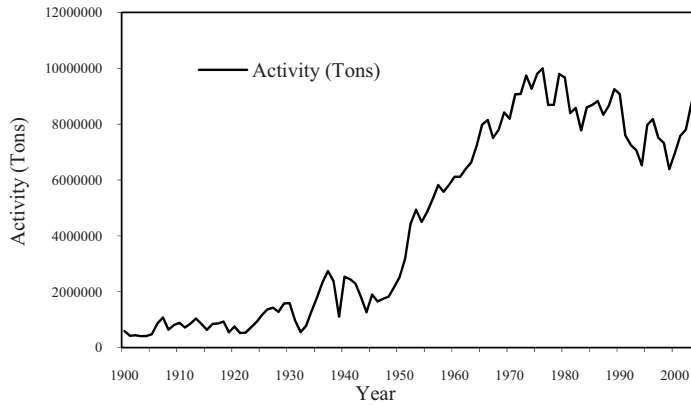


Figure 23. Manganese production activity covering the years 1900-2004. In 2004 Mn was in Stage IV, or the survival stage, with production oscillating and not rising or dropping permanently over time. 2004 production was 9,350,000 tons and not above the peak production of 10,000 tons in 1976.

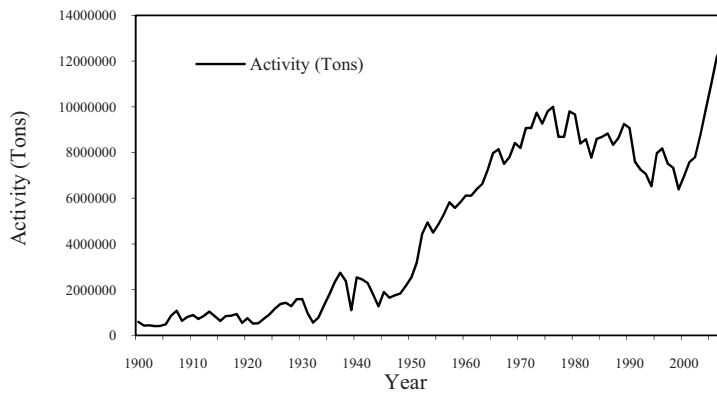


Figure 24. Manganese production activity covering the years 1900-2007. In 2007 Mn is in Stage III. Oscillations in production have stopped with a continual rise in production from 1999 –2007. 2007 production was 12,600,000 tons. 1976 production was surpassed in 2005.

Figure 24 shows manganese production for the years 1900-2007. Indicated is a steady continuous rise in production from 1976 to 2007. The peak production of 1976 was surpassed in 2005 with increases until 2007 where Mn re-entered Stage III. Production plots graphically illustrate Stage III features and Stage III is possibly further indicated by the presence of an origin shift and a linear alpha in its best-fit equation rather than a parabolic alpha, which is

required for Stage IV materials [2]. It is possible that in another span of three years these materials could return to Stage IV due to decline in production or oscillation caused by resources depletion or economic and environmental forces. If there is a steady rise in the production data over the last few years evaluated, the material or system will likely be in Stage III. If there is steady decline or no change in production data over the last few years of production, Stage IV is probably present. The stage depends on whether the material or system is in the peak or valley of the oscillating production data at the most recent date.

Relevance

The processes and evaluations presented in this dissertation set forth original and important tools that can possibly be used for the prediction of the innovative behavior of engineering materials and energy sources. Each step in the evaluation process presented here provides a certain level of predictive ability for the more efficient usage of ever more scarce natural resources. Also, the use of these procedures allow for the wiser and more profitable allocation of monetary resources that in light of the present economic environment, are vital for financial stability for individuals, corporations and nations.

Correlation theory provides the first predictive tool. As has been demonstrated, resources such as engineering materials and energy sources in almost every case tested, display some degree of correlation. This means a change in one of the data sets, such as production, was statistically the result of changes in the patent data set or vice versa. From this, it can be suggested that if an individual or group has information that reveals probable future increases in production, such as governmental stimulus monies available for production of energy sources, it may be profitable to participate in future innovation, since, if correlation is high for the material or system the patenting curve will follow the trend of the production curve. Likewise, if it is announced that governmental funds will be available for

innovative activity applied to renewable energy it could signal an opportunity to provide resources for the production of renewable energy if there is strong correlation since the increase in production will mirror the increase in innovation. If production is predicted to drop in the future due to resource depletion, environmental concerns or global conditions it can be inferred that innovation will decline as well and decisions concerning development can be based upon this knowledge.. The stronger that the correlation between the production and patenting data of a material or system is, will make any development or research decisions more logical and intelligent.

Best-fit analysis and the identification of the long-term life cycles of materials and system also provide tools in the allocation of natural and economic resources. It has previously been proven that metals and non-metals have four-stage life cycles, and above, it has been shown that energy sources also have a four-stage life cycle similar to metals and non-metals allowing for identical evaluation. The common pattern equation (Eq. 1) and best-fit analysis can be used to show the life cycle of engineering materials and energy sources by the plots generated as well as the R^2 values calculated. Determining the likely life cycle point that a material or system is in currently allows for intelligent development and allocation decisions. Development of a material or system found to be in Stage III of its production life cycle would be a more logical choice than development of a Stage IV material. Stage III materials are in a time of rapid growth while those in Stage IV are in survival mode with little growth. If growth occurs in a Stage IV it is probably for the short term due to the oscillations in activity existing in this stage. A Stage IV material may return to Stage III or it may fall into a possible Stage V. There is thus far more risk in allocating resources in Stage IV materials than in Stage III. However, Stage IV materials often offer opportunities in the innovation around or in replacement of them, which again is an example of the destructive side of the innovative process.

Indicators of a Stage III engineering material and energy source are firstly strong correlation between the data sets. Other indications of Stage III are R^2 values near one for engineering materials and positive for energy sources in regards to production activity and patenting best-fit evaluation, while the presence of an origin shift is final evidence of the presence of Stage III. It is suggested here that the presence of an origin shift in the patent data of a material confirms the presence of Stage III that is predicted by the previously mentioned indicators. In all cases for engineering materials and energy sources, the models proposed in this dissertation have shown that when Stage III seems to be likely, that an origin shift is in existence.

The origin shifts produced by best-fit analysis offer a possible time frame in which the allocation of monetary resources for research and development may more safely take place. If a material or system experiences a negative origin shift in patent best-fit, its patenting or innovation occurs before its production at an identical point in their respective life cycles. This suggests that if a material or system is in Stage III and the innovative activity happens X number of years previous to the production, then when the production life cycle curve enters Stage IV, the innovative activity won't reach that same point in its life cycle for that X number of years. This outcome possibly gives a window of X number of years to consider innovation relating to that material.

On the other hand if there is a positive shift in the patent best-fit origin the innovative activity reaches a point in its life cycle after the production activity crosses the same point in its life cycle. It may be less profitable to invest time and money on a material or system if this is the case. Consideration may be needed to spend less on the production for a material whose innovative activity has reached Stage IV a number of years before its production has. If innovation is declining there may be less of a reason to continue further production.

U.S. wind power, for example, has a strong correlation coefficient of 0.9681, a high best-fit R^2 of 0.8561, is likely in Stage III and has a negative origin shift of 22 years. Development of wind power is likely a good decision. It is in Stage III, or rapid growth stage and has production that is being driven by its innovation, where the constructive side of the innovative process exists and the universal constant, n_0 , and drive ratio are below one, as indicted by the negative origin shift in patent data. Twenty-two years before the production data of wind power reached a specific point in its life cycle, its patent data reached that same point. In other words, 22 years after the production of wind power reaches the survival stage (Stage IV), the innovative behavior, as measured by the patent data will reach a like point in its life cycle. Such gives a possible 22 year window to safely allocate resources for the development of wind power after it has reached Stage IV in its production.

The evaluation of kyanite, on the other hand, produced a positive 20 year shift in its patent data origin indicating that innovative activity in this case is being driven by production. Twenty years after the production curve reaches a certain point in its life cycle the patenting data will reach the same point in its life cycle. The destructive side of the innovation process is present and innovation will still be present but to a lesser degree than where constructive innovation exists. With kyanite, or other materials or systems with positive origin shifts, an individual, company or nation might want to be more cautious about allocating resources since there will be less innovative activity and the production will reach Stage IV before the patenting data. More information would be needed to make an intelligent choice. If it is foreseen, with kyanite, that its production Stage IV will last 20 years or longer it may still be profitable to invest resources in it. In such a case opportunities will be present in innovating around the technology or in replacement of it or to use present production.

It is strongly believed that presented above is a powerful method for the evaluation of the relationship of engineering materials and energy sources, which can be employed, for the

allocation of scarce economic and natural resources. Varying fields such as engineering, economics, law and business will all benefit from the predictive value that this model provides.

Conclusion and Summary: The study presented in this paper has shown that patents are a good measure of technical innovation for engineering materials and energy sources. For Stage III materials and energy sources, the patents mirror and behave similarly to the production activity as described by the life cycle platform (1-3,83). Correlation theory and best-fit analysis when applied to the fifty engineering materials and fourteen energy sources studied provide several indicators that when combined can suggest the stage of a material or system and the driving force of the innovative activity of the material or system. Stage III is generally indicated by a life cycle curve that is still growing, strong correlation and a production and patenting data R^2 values near one. The existence of an origin shift verifies the existence of Stage III for engineering materials and energy sources. Even more, a negative origin shift and low drive ratio points towards the strongest innovative activity for a Stage III material or system, while a positive shift and higher ratio indicates not as strong innovative activity.

Although not conclusively established as the sole reason, the dual nature of Schumpeterian innovation is possibly also highlighted when comparing the patents and production of the various substances. It has been found that the life cycle stage of a material or energy source, as well as the relative extent of the driving force acting upon that item's production can be determined. When the drive ratio of a material or system is below the universal constant n_0 , of one, the material or substance may be in the creative mode of innovation where patents and innovation spur economic growth. When the drive ratio is greater than one, or n_0 , the destructive side of the innovative process might be present. The y-

axis of the plots shown in Figs. 12(a-c) and 20, which correlate the driving force to origin shift, appears to always be at 1 when the origin shift is 0, thus perhaps indicating that there is one universal constant, common to the life cycle of all engineering materials and energy sources.

Appendix 1: Correlation

The correlation, or sample correlation coefficient, r , is calculated with the equation,

$$r = S_{xy} / (S_{xx} * S_{yy})^{1/2} \quad (A1.1),$$

where

$$S_{xx} = \sum_{i=1}^n x_i^2 - (\sum_{i=1}^n x_i)^2 / n \quad (A1.2)$$

$$S_{yy} = \sum_{i=1}^n y_i^2 - (\sum_{i=1}^n y_i)^2 / n \quad (A1.3)$$

$$S_{xy} = \sum_{i=1}^n x_i y_i - (\sum_{i=1}^n x_i) (\sum_{i=1}^n y_i) / n \quad (A1.4)$$

and n is equal to the number of pairs of x and y in the data set. As an example, the correlation coefficient will be determined for the accumulated data concerning aluminum. In the case of aluminum, 108 data entries were made for activity and patents, x and y respectively, giving an n value of 108. For aluminum,

$$\Sigma x = 833134000, \Sigma y = 486253, \Sigma x^2 = 1.6e16, \Sigma y^2 = 5.81e9 \text{ and } \Sigma xy = 9.43e12.$$

Calculating, using equations A1.2, A1.3 and A1.4 and the above values,

$$S_{xx} = 16e16 - (833134000)^2 / 108 = 9.559e15,$$

$$S_{yy} = 5.81e9 - (486253)^2 / 108 = 3.62e9 \text{ and}$$

$$S_{xy} = 9.43e12 - \{833134000\} * \{486253\} / 108 = 5.68e12.$$

By substitution of the above values into the Eq. (A1.1), $r = S_{xy} / (S_{xx} * S_{yy})^{1/2}$

we get

$$5.68e12 / (9.559e15 * 3.62e9)^{1/2} = 0.965222$$

which indicates a good correlation between the activity and patent data. The r value of 0.9652 can be squared then multiplied by 100 resulting in 93.17 suggesting that 93.17% of the variations in the patent numbers can be attributed to corresponding differences in the activity data [81-83]. Table A1.1 below displays the results calculated above and Fig. A1.1 shows a comparison of aluminum activity and patent data. Each material was treated in the same manner.

Table A1.1. Correlation Eq.(A1.1) terms calculated from Table A3.1 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
833134000	486253	1.6E+16	5.81E+09	9.43E+12	9.559E+15	3.62E+09	5.68E+12	0.965222	93.16534

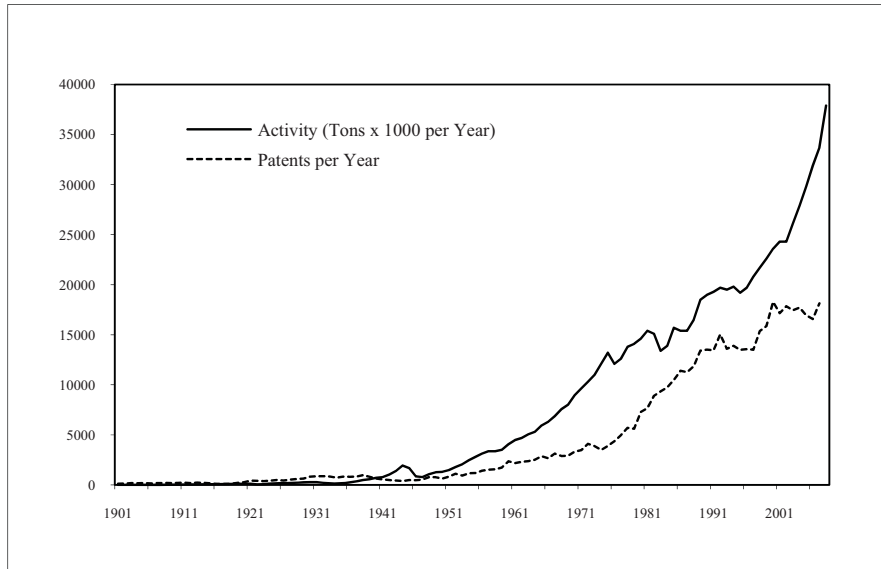


Figure A1.1. Aluminum activity and patent data illustrating correlation.

Appendix 2: MatLab

Program A2.1: MatLab program template used for Best-Fit analysis [2]. Common pattern equation (1) is contained within the program. Individual parameters, x (activity), y and r (years) entered for each material. A plot with individual curves, displaying the actual activity values and fitted data are produced as well as an R^2 value is produced by this program.

```
clear all;
%% %START USER INPUTS
%~~~~~
year=[];%please input the years in the brackets. separate individual values by a comma and separate lines by
'.....'
    %example [1990, 1991.....
    %      1992]
activity=[];%please input the years in the brackets. separate individual values by a comma and separate lines
by '.....'
    %example [2.8, 3.6.....
    %      4.8]

%constants
%Please choose the values for the parameters below
%~~~~~
a= ;    %alpha
b= ;    %Beta
u= ;    %mu "u" is the location parameter. It gives the position of the mean
v= ;    %nu "v" is the scale factor, which controls the width of peak
const= ; %delta
c= ;    %omega
x0=;    %origin of the data
n= ;    %power exponent

r= [];  %choose years where you want to predict
        %syntax: (1) r=<start year>:<increment>:<end year>;
        %      (2) r=[1900, 1901, 1903, ....,2000];

%% %END USER INPUTS
%~~~~~

% From data
plot(year,activity*1e3,'r'); %plots (Year VS Activity)

hold on; %to plot two graphs in the same figure

% From Equation
x=r-x0; %normalized years
R2=0; %non-linear R^2 value intialized to zero

% EQUATIONS
z=const*(exp((x-u)./v).*exp(-exp((x-u)./v)))/v;
%calculation of the additive part in the equation
```

```

y=x.^n.*[(a^n.*x.^2)+(b^n*sin(c*x))]+z;
%calculation of the complete equation

plot(r,y,'b');
%plotting the equation

xlabel('Years'); ylabel('Activity');

%Calculation of R2
actmean=sum(y)/length(y);

for i=1:length(y)
    SStot=(activity*1e3-actmean).^2;
    SSreg=(y-activity*1e3).^2;
end
R2=1-(sum(SSreg)/sum(SStot));
disp(R2);
%displaying the R^2 value

```

Program A2.2: Zinc Production MatLab Program. The years 1900-2007 are entered for the year and “r” inputs while the production data for each year is entered in the activity input. The parameters a, b, u, v, const, c, x0 and n (α , β , μ , ν , δ , ω , x_0 and n respectively) are entered as well. Running the program results in a plot and an R^2 value. The best-fit is the curve with an R^2 value closest to one indicating the best estimate of the equation parameters [2]. Each material was evaluated in the same manner.

```

clear all;
% %START USER INPUTS
% ~~~~~
year=[1900, 1901, 1902, 1903, 1904, 1905, 1906, 1907, 1908, 1909, 1910, 1911, 1912, 1913, 1914, 1915,
1916, 1917, 1918, 1919, 1920.....
    1921, 1922, 1923, 1924, 1925, 1926, 1927, 1928, 1929, 1930, 1931, 1932, 1933, 1934, 1935, 1936, 1937,
1938, 1939, 1940.....
    1941, 1942, 1943, 1944, 1945, 1946, 1947, 1948, 1949, 1950, 1951, 1952, 1953, 1954, 1955, 1956, 1957,
1958, 1959, 1960.....
    1961, 1962, 1963, 1964, 1965, 1966, 1967, 1968, 1969, 1970, 1971, 1972, 1973, 1974, 1975, 1976, 1977,
1978, 1979, 1980.....
    1981, 1982, 1983, 1984, 1985, 1986, 1987, 1988, 1989, 1990, 1991, 1992, 1993, 1994, 1995, 1996, 1997,
1998, 1999, 2000.....
    2001, 2002, 2003, 2004, 2005, 2006, 2007];%please input the years in the brackets. separate individual
values by a comma and separate lines by '.....'
    %example [1990, 1991.....
    %    1992]
activity=[479, 510, 547, 574, 629, 660, 704, 738, 723, 775, 810, 895, 971, 939, 795, 760, 882, 901, 849, 719,
682, 464, 730.....
    889, 986, 1190, 1410, 1420, 1360, 1320, 1260, 904, 709, 892, 1060, 1210, 1330, 1470, 1420, 1500, 1470,
1590, 1630, 1830, 1870.....
    1470, 1440, 1600, 1690, 1730, 2150, 2360, 2590, 2670, 2660, 2900, 3110, 3150, 2950, 3020, 3090, 3490,
3570, 3660, 4030, 4310.....
    4500, 4840, 4970, 5340, 5460, 5520, 5440, 5710, 5780, 5850, 5690, 5920, 5850, 5990, 5950, 5950, 6130,
6280, 6520, 6760, 6840, 7190.....
    6770, 6820, 7150, 7270, 7250, 6910, 7050, 7280, 7480, 7540, 7570, 7960, 8770, 8910, 8880, 9520, 9590,
9930, 10000, 10900];%please input the years in the brackets. separate individual values by a comma and
separate lines by '.....'

```

```

%example [2.8, 3.6.....
%      4.8] 86

%constant
%Please choose the values for the parameters below
%-----
a= 20;      %alpha
b= 30;      %Beta
u= 27;      %mu "u" is the location parameter. It gives the position of the mean
v= 0.9;     %nu "v" is the scale factor, which controls the width of peak
const= 1.5e6; %delta
c= 0.7;     %omega
x0= 1900;   %origin of the data
n= 0.9;     %power exponent

r= [1900, 1901, 1902, 1903, 1904, 1905, 1906, 1907, 1908, 1909, 1910, 1911, 1912, 1913, 1914, 1915, 1916,
1917, 1918, 1919, 1920.....
    1921, 1922, 1923, 1924, 1925, 1926, 1927, 1928, 1929, 1930, 1931, 1932, 1933, 1934, 1935, 1936, 1937,
1938, 1939, 1940.....
    1941, 1942, 1943, 1944, 1945, 1946, 1947, 1948, 1949, 1950, 1951, 1952, 1953, 1954, 1955, 1956, 1957,
1958, 1959, 1960.....
    1961, 1962, 1963, 1964, 1965, 1966, 1967, 1968, 1969, 1970, 1971, 1972, 1973, 1974, 1975, 1976, 1977,
1978, 1979, 1980.....
    1981, 1982, 1983, 1984, 1985, 1986, 1987, 1988, 1989, 1990, 1991, 1992, 1993, 1994, 1995, 1996, 1997,
1998, 1999, 2000.....
    2001, 2002, 2003, 2004, 2005, 2006, 2007] %choose years where you want to predict
%syntax: (1) r=<start year>:<increment>:<end year>;
%      (2) r=[1900, 1901, 1903, ....,2000];

% %END USER INPUTS
%-----

% From data
plot(year,activity.*1e3,'r'); %plots (Year VS Activity)

hold on; %to plot two graphs in the same figure

% From Equation
x=r-x0; %normalized years
R2=0; %non-linear R^2 value intialized to zero

% EQUATIONS
z=const*(exp((x-u)./v).*exp(-exp((x-u)./v)))./v;
%calculation of the additive part in the equation

y=x.^n.*[(a^n.*x.^2)+(b^n.*x.*sin(c*x))]+z;
%calculation of the complete equation

plot(r,y,'b');
%plotting the equation

xlabel('Years'); ylabel('Activity');

%Calculation of R2
actmeany=sum(y)/length(y);

```



```

for i=1:length(y)
    SStot=(activity*1e3-actmeany).^2;
    SSreg=(y-activity*1e3).^2;
end
R2=1-(sum(SSreg)/sum(SStot));
disp(R2);
%displaying the R^2 value

```

Program A2.3: Data resulting from program A2.2 composed of “r” values, an R^2 of 0.8805 and an actual and best-fit curve plot.

r =

Columns 1 through 9

1900	1901	1902	1903	1904	1905	1906	1907	1908
------	------	------	------	------	------	------	------	------

Columns 10 through 18

1909	1910	1911	1912	1913	1914	1915	1916	1917
------	------	------	------	------	------	------	------	------

Columns 19 through 27

1918	1919	1920	1921	1922	1923	1924	1925	1926
------	------	------	------	------	------	------	------	------

Columns 28 through 36

1927	1928	1929	1930	1931	1932	1933	1934	1935
------	------	------	------	------	------	------	------	------

Columns 37 through 45

1936	1937	1938	1939	1940	1941	1942	1943	1944
------	------	------	------	------	------	------	------	------

Columns 46 through 54

1945	1946	1947	1948	1949	1950	1951	1952	1953
------	------	------	------	------	------	------	------	------

Columns 55 through 63

1954	1955	1956	1957	1958	1959	1960	1961	1962
------	------	------	------	------	------	------	------	------

Columns 64 through 72

1963	1964	1965	1966	1967	1968	1969	1970	1971
------	------	------	------	------	------	------	------	------

Columns 73 through 81

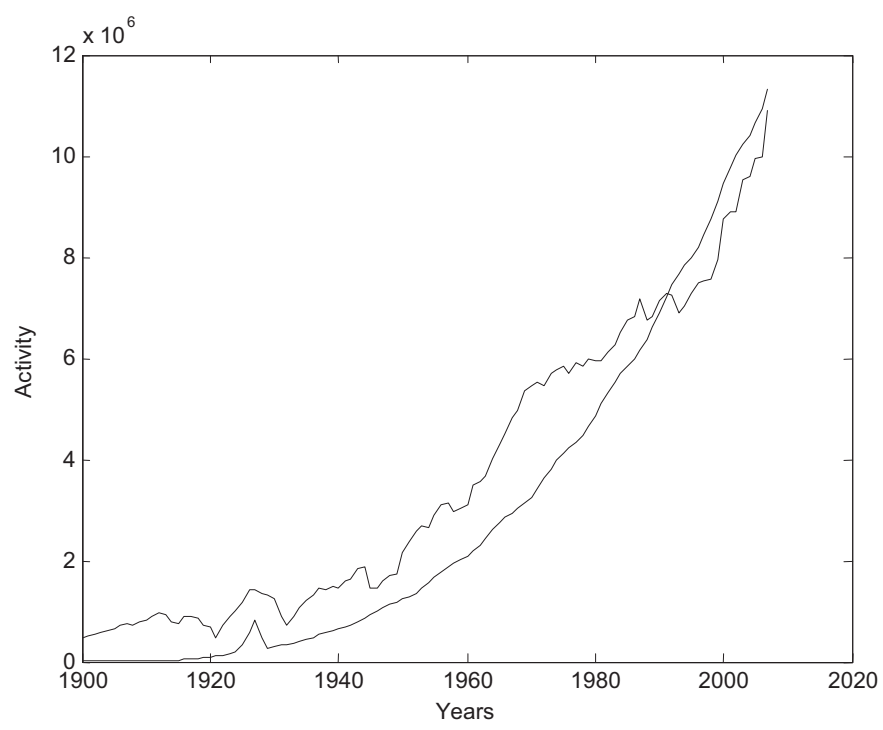
1972	1973	1974	1975	1976	1977	1978	1979	1980
------	------	------	------	------	------	------	------	------

Columns 82 through 90

1981	1982	1983	1984	1985	1986	1987	1988	1989
------	------	------	------	------	------	------	------	------

Columns 91 through 99

1990 1991 1992 1993 1994 1995 1996 1997 1998
Columns 100 through 108
1999 2000 2001 2002 2003 2004 2005 2006 2007
0.8805



Appendix 3: Engineering Material Data

Table A3.1. Aluminum Activity³ and Patents⁴

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	6800	117	1927	220000	539	1954	2810000	1170	1981	15100000	7673
1901	6800	129	1928	258000	592	1955	3140000	1189	1982	13400000	8912
1902	7900	143	1929	280000	644	1956	3370000	1420	1983	13900000	9348
1903	8500	187	1930	272000	841	1957	3370000	1531	1984	15700000	9747
1904	10000	177	1931	220000	868	1958	3510000	1567	1985	15400000	10485
1905	13000	187	1932	153000	876	1959	4060000	1729	1986	15400000	11407
1906	17000	155	1933	142000	848	1960	4490000	2372	1987	16500000	11271
1907	22000	181	1934	170000	735	1961	4700000	2181	1988	18500000	11856
1908	17000	207	1935	259000	832	1962	5060000	2300	1989	19000000	13416
1909	30000	187	1936	360000	824	1963	5320000	2379	1990	19300000	13523
1910	45000	197	1937	482000	827	1964	5940000	2519	1991	19700000	13455
1911	46000	233	1938	579000	984	1965	6310000	2871	1992	19500000	15034
1912	58000	192	1939	720000	823	1966	6880000	2668	1993	19800000	13599
1913	65000	233	1940	787000	635	1967	7570000	3129	1994	19200000	13897
1914	69000	213	1941	1040000	563	1968	8020000	2895	1995	19700000	13498
1915	78000	159	1942	1400000	491	1969	8970000	2917	1996	20800000	13572
1916	106000	116	1943	1950000	436	1970	9650000	3325	1997	21700000	13505
1917	123000	117	1944	1690000	397	1971	10300000	3480	1998	22600000	15369
1918	128000	119	1945	870000	486	1972	11000000	4112	1999	23600000	15872
1919	121000	199	1946	790000	473	1973	12100000	3884	2000	24300000	18286
1920	125000	279	1947	1080000	548	1974	13200000	3496	2001	24300000	17164
1921	70000	429	1948	1270000	784	1975	12100000	3908	2002	26100000	17853
1922	87000	409	1949	1310000	776	1976	12600000	4373	2003	27900000	17453
1923	141000	389	1950	1490000	637	1977	13800000	4961	2004	29800000	17721
1924	168000	435	1951	1800000	837	1978	14100000	5705	2005	31900000	16944
1925	178000	501	1952	2060000	1124	1979	14600000	5620	2006	33700000	16577
1926	195000	449	1953	2470000	936	1980	15400000	7279	2007	37900000	18141

Table A3.2. Correlation Eq.(A1.1) terms calculated from Table A3.1 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
833134000	486253	1.6E+16	5.81E+09	9.43E+12	9.559E+15	3.62E+09	5.68E+12	0.965222	93.16534

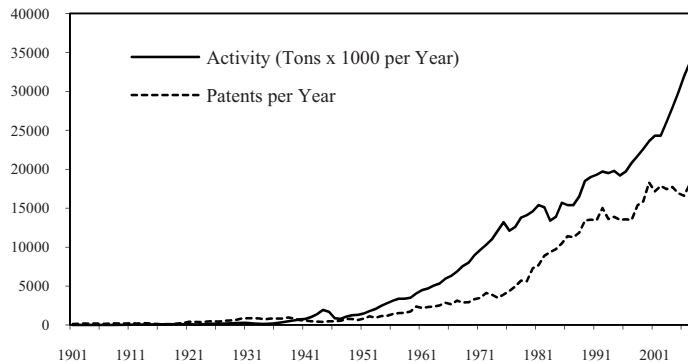


Figure A3.1. Aluminum: Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

³ Activity represents world production of aluminum, defined at usgs.gov as "...world primary aluminum production. Data are reported in the MR [Mineral Resources of the United States] and the MYB [Minerals Yearbook]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Aluminum, Al and aluminium were used as keywords found in the patent title or abstract by year of publication.

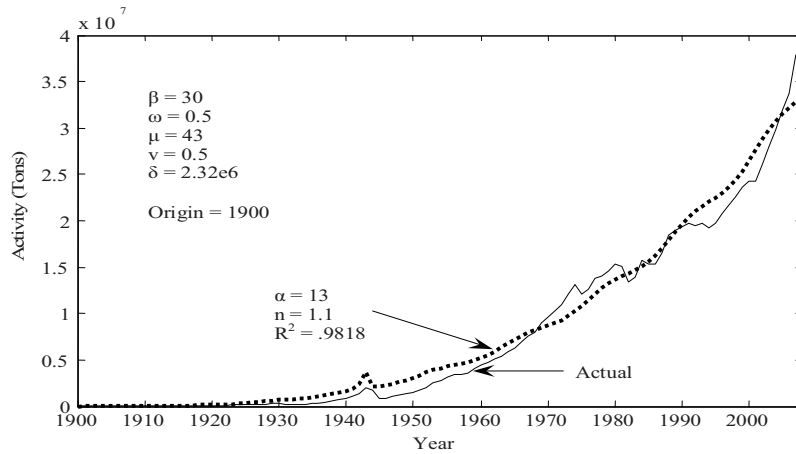


Figure A3.2. USGS World Aluminum Production. World aluminum production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

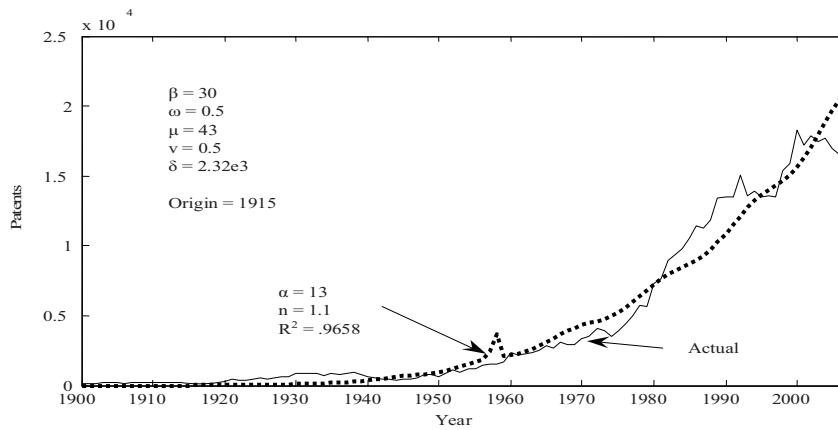


Figure A3.3. EPO Worldwide Patent Search: Aluminum, Al or Aluminum in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

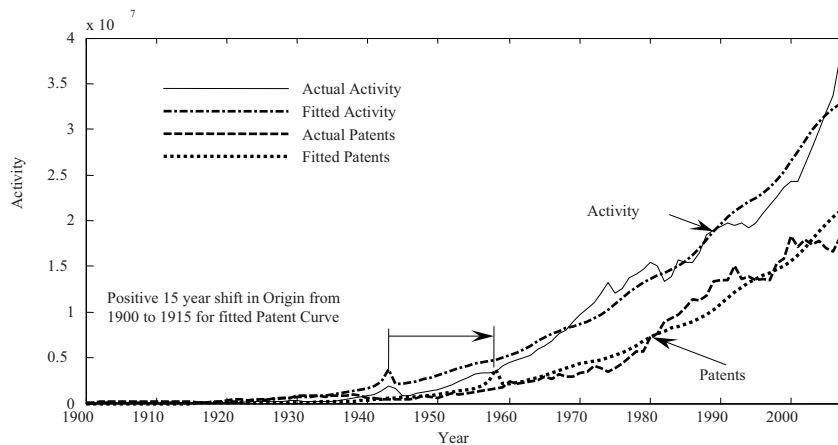


Figure A3.4. Aluminum Best-Fit Activity and Patents. Illustrates aluminum best-fit origin shift.

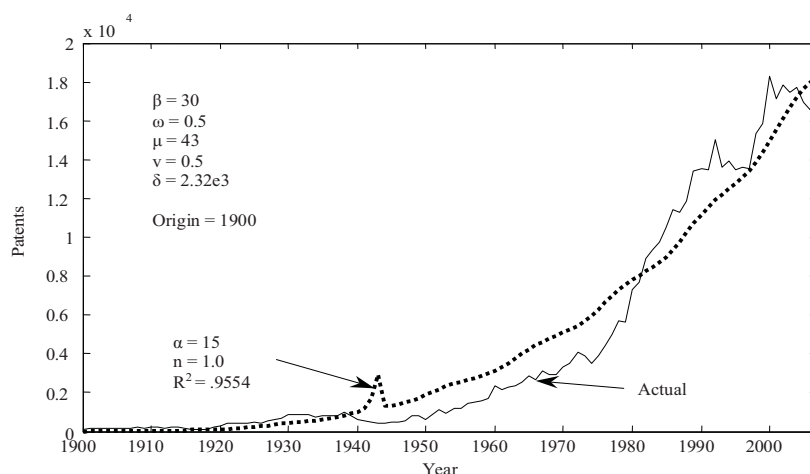


Figure A3.5. Aluminum Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.3. Antimony Activity⁵ and Patents⁶

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	28000	50	1954	39900	103	1981	59200	350
1901			1928	28500	51	1955	46300	83	1982	53800	366
1902			1929	31600	64	1956	53300	130	1983	48400	373
1903			1930	23600	86	1957	50800	116	1984	53400	401
1904			1931	15600	82	1958	46300	139	1985	55000	443
1905			1932	17300	67	1959	53300	119	1986	59900	401
1906	14500	19	1933	20200	68	1960	53300	225	1987	56100	367
1907	15000	17	1934	22600	61	1961	51900	192	1988	64400	439
1908	16000	15	1935	29800	64	1962	53700	201	1989	68400	479
1909	15000	15	1936	35300	78	1963	58000	227	1990	60400	496
1910	15000	22	1937	38600	101	1964	63000	254	1991	64700	515
1911	15500	34	1938	33900	79	1965	63000	313	1992	76000	518
1912	24200	26	1939	38800	50	1966	61400	217	1993	73000	509
1913	24500	22	1940	46300	66	1967	58400	243	1994	106000	537
1914	23600	21	1941	49000	51	1968	61500	220	1995	103000	487
1915	43200	13	1942	51400	36	1969	66200	198	1996	156000	504
1916	81600	8	1943	53200	32	1970	70000	187	1997	155000	530
1917	57200	13	1944	36000	27	1971	64100	208	1998	117000	548
1918	30800	18	1945	27000	52	1972	68100	243	1999	108000	597
1919	11800	13	1946	26000	56	1973	69300	209	2000	118000	625
1920	29000	19	1947	38000	42	1974	70500	209	2001	157000	656
1921	18300	27	1948	45000	64	1975	67900	262	2002	118000	778
1922	18900	31	1949	37000	64	1976	69200	305	2003	116000	700
1923	17600	30	1950	50000	57	1977	72200	296	2004	142000	649
1924	17500	33	1951	65000	57	1978	68800	279	2005	171000	618
1925	25500	44	1952	44500	86	1979	71900	268	2006	173000	586
1926	29000	51	1953	33600	63	1980	67200	371	2007	170000	621

⁵ Activity represents world production of antimony, defined at usgs.gov as "...world mine production in terms of antimony content. U.S. production is withheld and not available in the total for the years 1985-92 and 2000 to the most recent. Data are reported in the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁶ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Antimony and stibium were used as keywords found in the patent title or abstract by year of publication.

Table A3.4. Correlation Eq.(A1.1) terms calculated from Table A3.3 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
5899990	22160	4.79E+11	9177362	1.94E+09	1.565E+11	4630458	7.25E+08	0.851817	72.55915

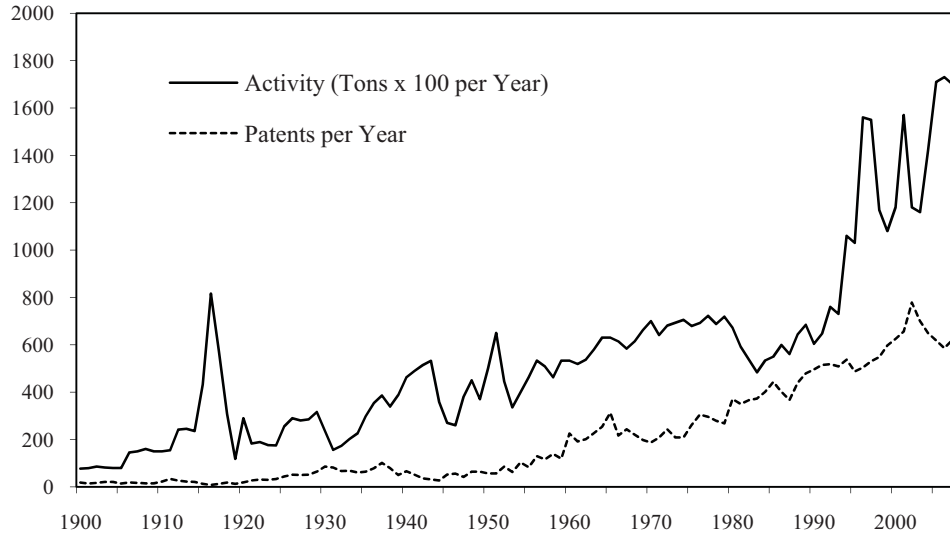


Figure A3.6. Antimony: Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

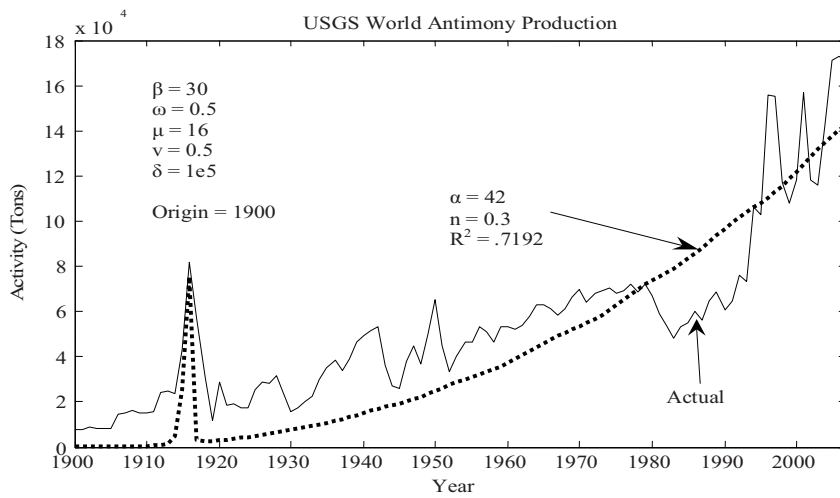


Figure A3.7. USGS World Antimony Production. World antimony production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

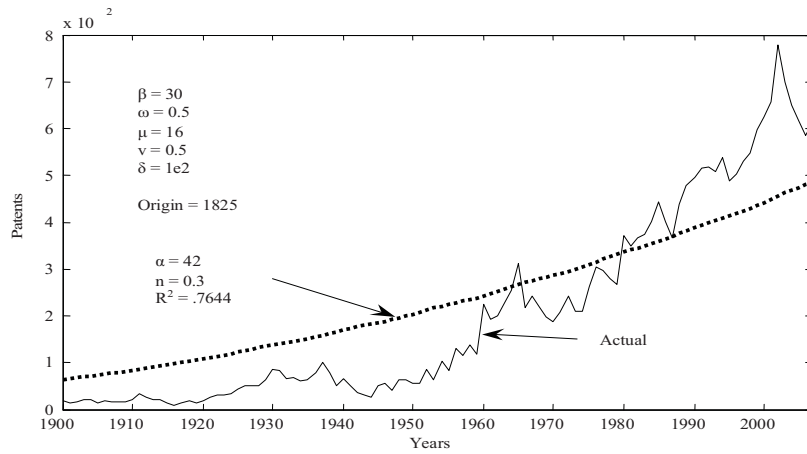


Figure A3.8. EPO Worldwide Patent Search: Antimony or Stibium in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

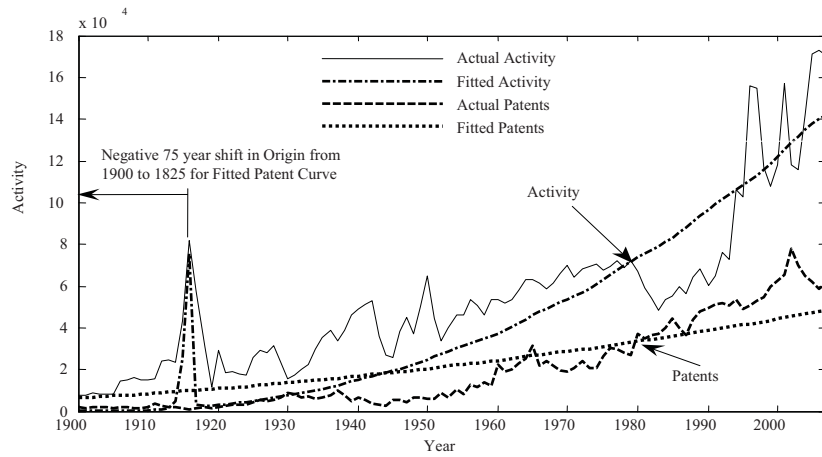


Figure A3.9. Antimony Best-Fit Activity and Patents. Illustrates antimony best-fit origin shift.

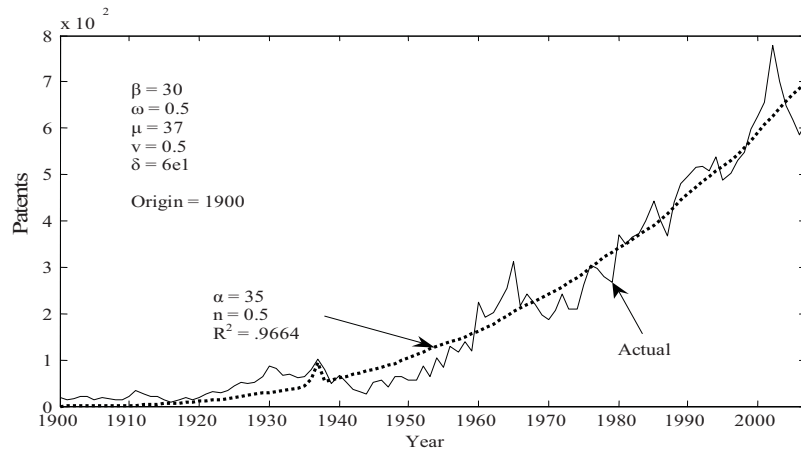


Figure A3.10. Antimony Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.5. Arsenic Activity⁷ and Patents⁸

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	39900	42	1954	26100	34	1981	33100	228
1901			1928	43200	58	1955	30900	41	1982	34500	281
1902			1929	46000	42	1956	33700	62	1983	31900	275
1903			1930	49600	59	1957	30900	62	1984	35600	334
1904			1931	51000	59	1958	27500	57	1985	40300	439
1905			1932	51500	45	1959	32100	54	1986	39600	435
1906			1933	27300	46	1960	39400	102	1987	47200	423
1907			1934	38200	40	1961	40500	93	1988	40400	500
1908			1935	41600	51	1962	34100	79	1989	49400	492
1909			1936	42200	42	1963	36600	121	1990	40400	436
1910	6810	17	1937	42400	57	1964	39900	103	1991	34800	457
1911	12800	12	1938	52200	44	1965	38600	135	1992	34700	497
1912	27600	17	1939	42800	42	1966	39500	117	1993	31900	349
1913	13400	13	1940	40400	32	1967	44600	115	1994	35400	321
1914	11500	6	1941	46000	34	1968	46400	104	1995	35600	287
1915	13300	5	1942	47700	14	1969	37700	108	1996	32500	237
1916	13300	8	1943	50200	19	1970	37500	105	1997	31800	253
1917	16400	3	1944	51800	8	1971	37800	100	1998	30500	358
1918	23200	11	1945	42100	37	1972	31400	143	1999	31600	365
1919	19000	15	1946	31800	32	1973	35200	111	2000	47500	406
1920	27900	12	1947	42400	20	1974	37200	94	2001	45000	370
1921	16200	15	1948	40900	34	1975	30700	105	2002	44700	387
1922	23500	20	1949	26500	26	1976	26100	134	2003	69700	364
1923	47300	30	1950	35700	20	1977	23200	119	2004	57800	421
1924	47200	26	1951	47400	19	1978	23300	182	2005	60000	461
1925	50300	37	1952	37100	41	1979	22400	127	2006	61200	391
1926	41100	46	1953	20600	23	1980	23600	240	2007	55900	451

Table A3.6. Correlation Eq.(A1.1) terms calculated from Table A3.5 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
3579210	14344	1.44E+11	4431828	5.84E+08	1.344E+10	2332335	59759598	0.337535	11.39299

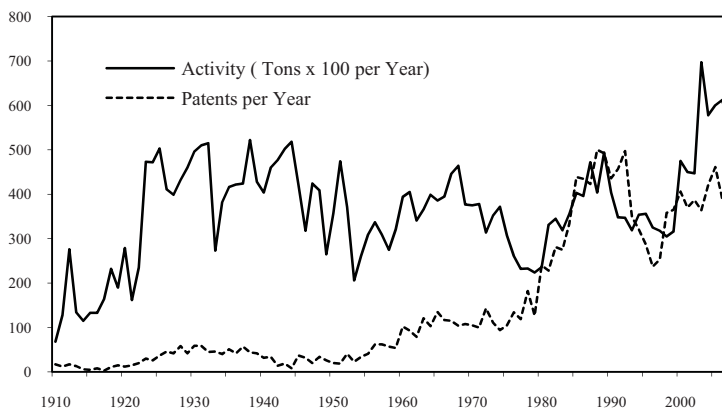


Figure A3.11. Arsenic: Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁷ Activity represents world production of antimony, defined at usgs.gov as “...world production of arsenic trioxide in terms of arsenic content. Data are not available for the years 1906-09. Data are from the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*].” Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁸ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Arsenic was used as a keyword found in the patent title or abstract by year of publication.

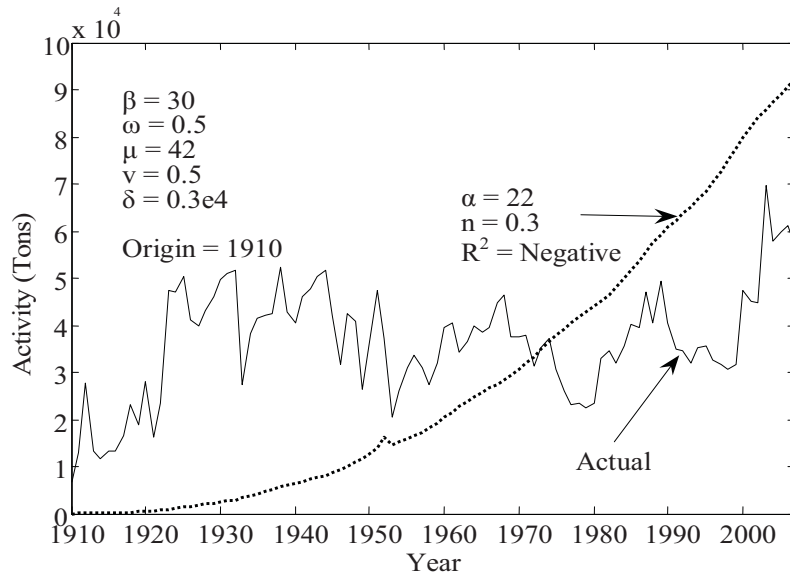


Figure A3.12. USGS World Arsenic Production. World arsenic production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. R^2 has a negative value indicating Stage IV. No best-fit for the patent data was obtainable.

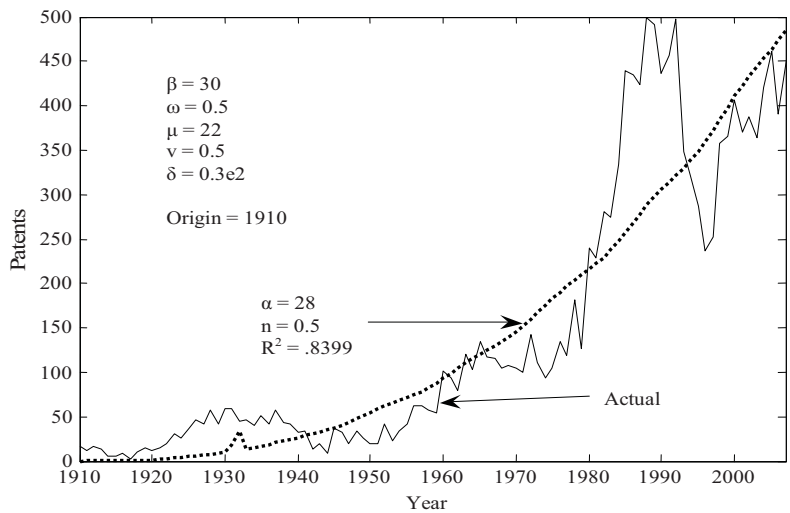


Figure A3.13. Arsenic Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.7. Asbestos⁹ Activity¹⁰ and Patents¹¹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	20600	104	1927	342000	209	1954	1510000	189	1981	4350000	324
1901	30500	117	1928	354000	189	1955	1770000	197	1982	4560000	325
1902	28400	154	1929	400000	205	1956	1810000	215	1983	4430000	313
1903	34300	183	1930	339000	245	1957	1890000	231	1984	4310000	272
1904	41100	180	1931	235000	238	1958	1860000	198	1985	4250000	254
1905	56000	154	1932	186000	276	1959	2050000	184	1986	4030000	213
1906	64700	126	1933	249000	232	1960	2210000	274	1987	4240000	203
1907	66300	133	1934	271000	240	1961	2510000	245	1988	4320000	220
1908	14100	146	1935	337000	229	1962	2410000	263	1989	4240000	337
1909	73600	130	1936	460000	207	1963	2510000	276	1990	4010000	324
1910	84700	135	1937	556000	247	1964	2770000	328	1991	3530000	304
1911	114000	150	1938	413000	246	1965	2810000	344	1992	3350000	333
1912	117000	156	1939	425000	182	1966	2970000	289	1993	2520000	269
1913	145000	166	1940	428000	155	1967	2910000	392	1994	2250000	221
1914	106000	153	1941	528000	144	1968	3010000	379	1995	2180000	201
1915	117000	128	1942	509000	131	1969	3270000	344	1996	2100000	207
1916	142000	91	1943	575000	118	1970	3490000	419	1997	2150000	153
1917	141000	78	1944	546000	94	1971	3580000	424	1998	1980000	192
1918	144000	94	1945	573000	127	1972	3780000	451	1999	1850000	149
1919	181000	136	1946	680000	144	1973	4190000	397	2000	2110000	157
1920	193000	168	1947	816000	155	1974	4160000	359	2001	2060000	111
1921	91100	203	1948	930000	224	1975	4140000	429	2002	2320000	141
1922	136000	198	1949	884000	209	1976	4770000	439	2003	2400000	140
1923	201000	171	1950	1290000	126	1977	4790000	444	2004	2330000	138
1924	198000	185	1951	1420000	157	1978	4690000	386	2005	2250000	126
1925	312000	195	1952	1420000	172	1979	4760000	249	2006	2180000	149
1926	329000	175	1953	1420000	173	1980	4700000	333	2007	2200000	349

Table A3.8. Correlation Eq.(A1.1) terms calculated from Table A3.7 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
187587400	23886	5.93E+14	6179482	5.28E+10	2.672E+14	896695	1.13E+10	0.728794	53.11403

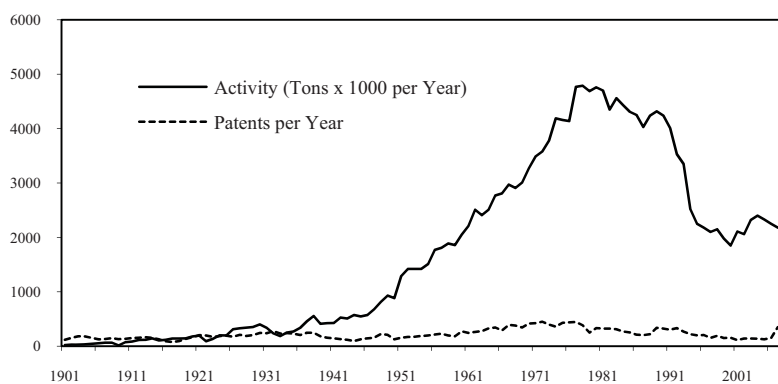


Figure A3.14 Asbestos: Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁹ Fibrous amphibole mineral [128].

¹⁰ Activity represents world production of asbestos, defined at usgs.gov as "...world mine production of asbestos. Data for the years 1900-04 include only Canada and the United States. For the years 1904-12 data include Canada, The United States and Russia. World asbestos mine production data are from the MR [Mineral Resources of the United States] and the MYB [Minerals Yearbook]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹¹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Asbestos was used as a keyword found in the patent title or abstract by year of publication.

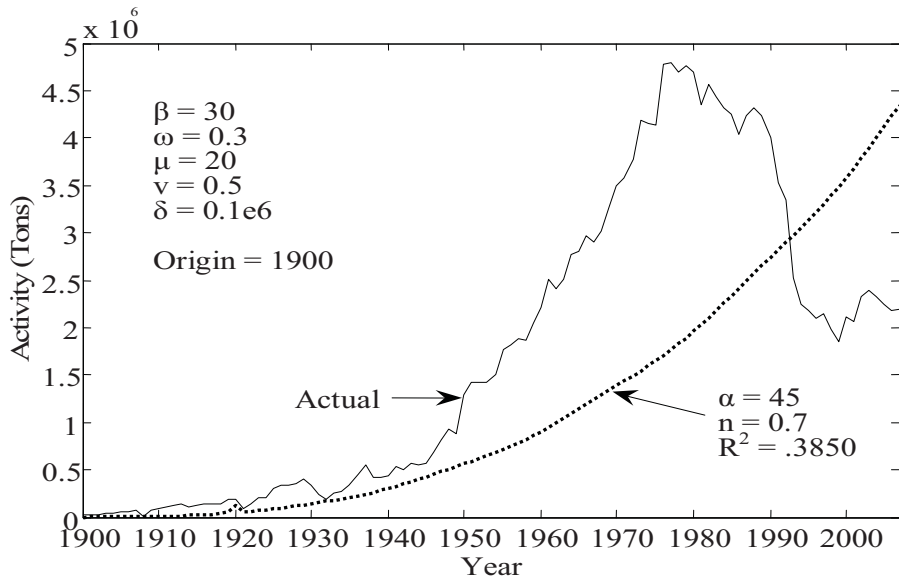


Figure A3.15. USGS World Asbestos Production. World asbestos production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

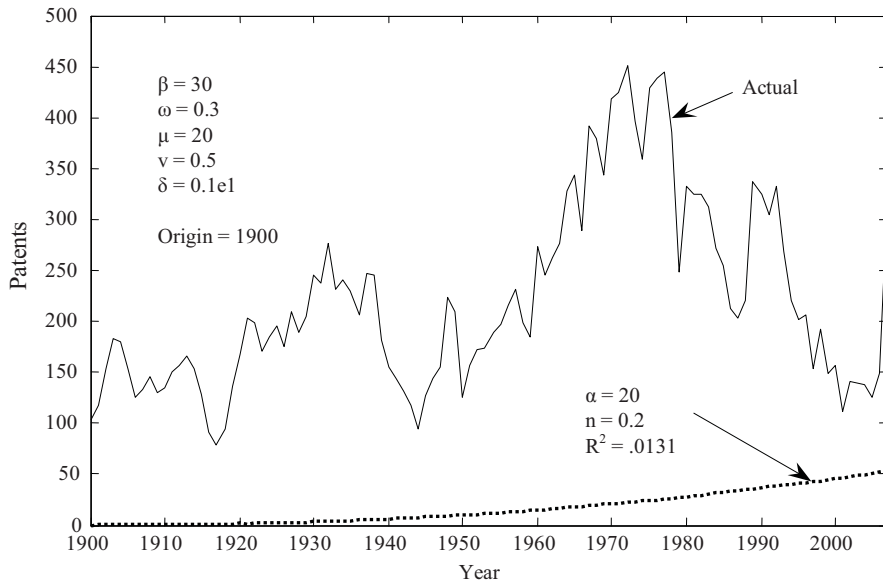


Figure A3.16. Asbestos Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.9. Barite¹² Activity¹³ and Patents¹⁴

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	536000	0	1954	2080000	4	1981	8310000	39
1901			1928	635000	0	1955	2420000	0	1982	7400000	45
1902			1929	677000	1	1956	2700000	1	1983	5470000	61
1903			1930	592000	0	1957	3340000	0	1984	5910000	72
1904			1931	468000	1	1958	2480000	1	1985	6160000	68
1905			1932	368000	0	1959	2700000	0	1986	4790000	92
1906			1933	439000	1	1960	2710000	0	1987	4810000	110
1907			1934	770000	0	1961	2820000	0	1988	5660000	64
1908			1935	726000	0	1962	3080000	1	1989	5820000	101
1909			1936	854000	0	1963	2880000	0	1990	5870000	95
1910			1937	992000	0	1964	3170000	1	1991	5570000	98
1911			1938	975000	1	1965	3540000	4	1992	4840000	105
1912			1939	961000	1	1966	3690000	2	1993	4470000	103
1913			1940	959000	0	1967	3570000	3	1994	4470000	139
1914			1941	1030000	0	1968	3420000	0	1995	4830000	120
1915			1942	1000000	1	1969	3850000	0	1996	6060000	126
1916			1943	1010000	0	1970	3940000	0	1997	6690000	135
1917			1944	1120000	1	1971	3730000	1	1998	5750000	126
1918			1945	964000	1	1972	3960000	10	1999	6160000	139
1919	359000	0	1946	1110000	0	1973	4750000	7	2000	6560000	157
1920	453000	0	1947	1360000	0	1974	4870000	5	2001	6740000	135
1921	242000	0	1948	1210000	0	1975	5010000	10	2002	6160000	128
1922	414000	1	1949	1340000	1	1976	5360000	13	2003	6780000	167
1923	445000	0	1950	1450000	0	1977	5960000	16	2004	7760000	149
1924	487000	2	1951	1670000	0	1978	7000000	18	2005	8110000	156
1925	536000	1	1952	1780000	1	1979	7170000	25	2006	7960000	133
1926	527000	0	1953	1960000	1	1980	7600000	52	2007	7630000	119

Table A3.10. Correlation Eq.(A1.1) terms calculated from Table A3.9 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
304959000	3171	1.59E+15	365171	1.96E+10	5.406E+14	252190.8	8.74E+09	0.748629	56.04449

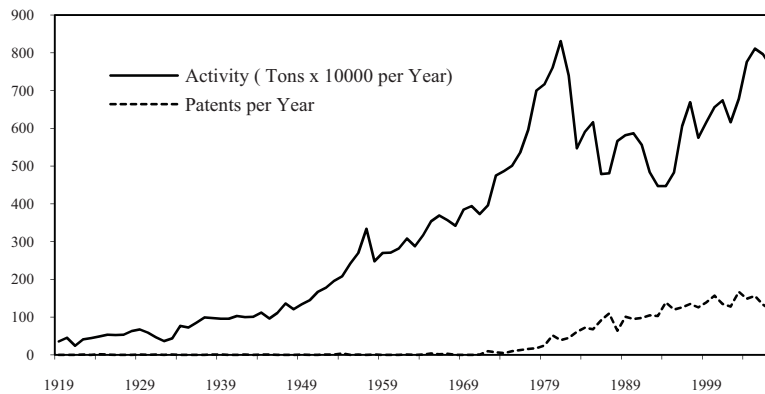


Figure A3.17. Barite: Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

¹² Orthorhombic mineral form of barium sulfate [128].

¹³ Activity represents world production of barite, defined at usgs.gov as "...world crude barite production. Data are not available for the years 1900-12 and 1914-1918. Data are reported in the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Barite, baryte or barium sulphate were used as keywords found in the patent title or abstract by year of publication.

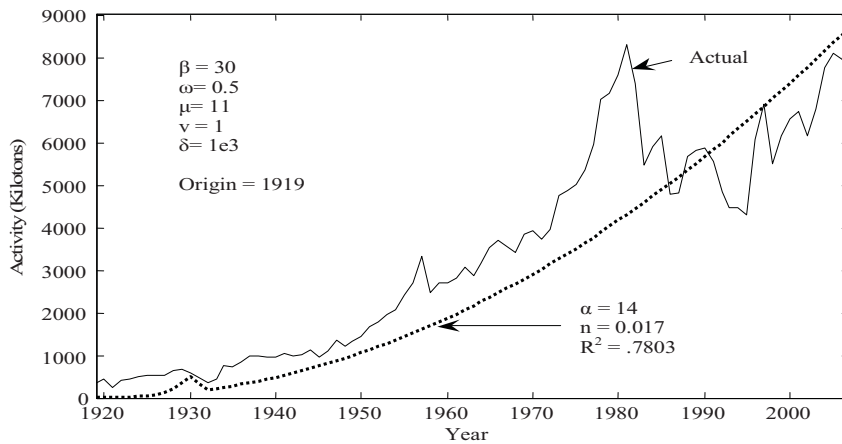


Figure A3.18. USGS World Barite Production. World barite production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

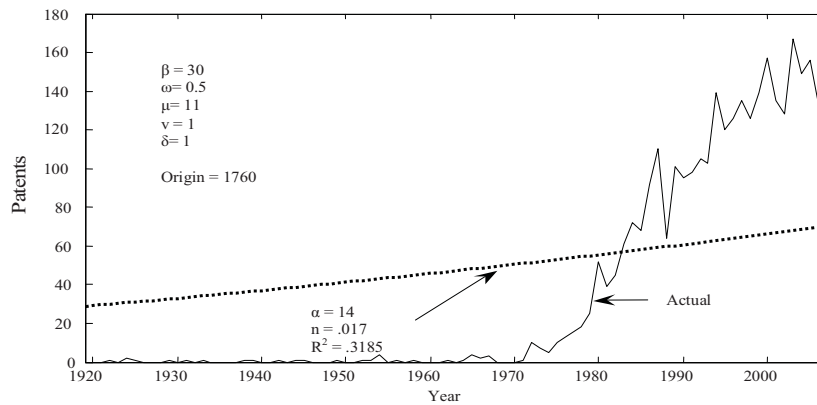


Figure A3.19. EPO Worldwide Patent Search: Barite, Baryte or Barium Sulphate in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

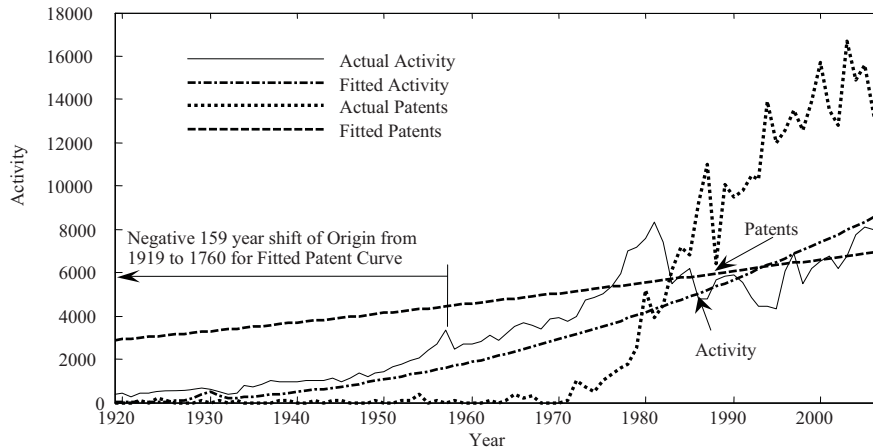


Figure A3.20. Barite Best-Fit Activity and Patents. Illustrates barite best-fit origin shift.

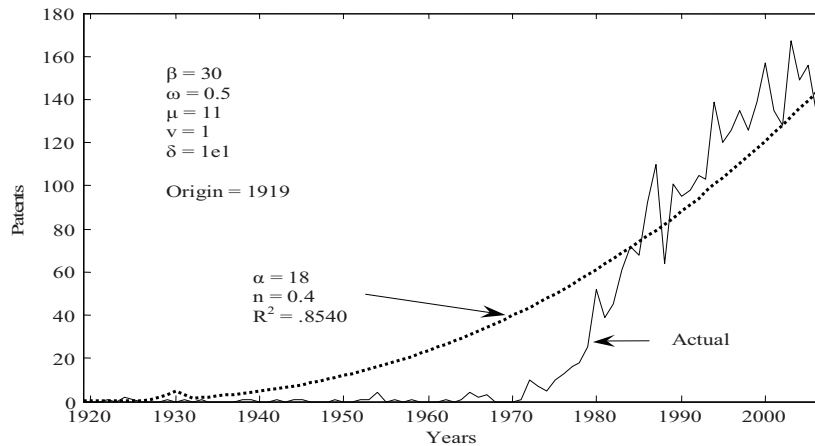


Figure A3.21. Barite Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.11. Bauxite/Alumina¹⁵ Activity¹⁶ and Patents¹⁷

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y patent	Yr.	x (activity)	y patent
1900	88000	19	1927	1880000	80	1954	16200000	278	1981	85300000	1466
1901	106000	21	1928	2030000	95	1955	17800000	260	1982	79300000	1605
1902	136000	16	1929	2150000	81	1956	18800000	331	1983	78700000	1674
1903	189000	26	1930	1630000	127	1957	20500000	374	1984	87200000	1854
1904	133000	27	1931	1150000	103	1958	21400000	432	1985	84200000	1895
1905	159000	19	1932	1000000	107	1959	23100000	414	1986	88200000	2162
1906	201000	20	1933	1100000	102	1960	27600000	581	1987	91600000	2004
1907	269000	22	1934	1330000	100	1961	29400000	434	1988	97400000	2207
1908	243000	37	1935	1770000	109	1962	31100000	435	1989	103000000	3058
1909	275000	28	1936	2830000	124	1963	30700000	453	1990	113000000	2794
1910	356000	27	1937	3750000	113	1964	33400000	502	1991	111000000	2652
1911	425000	23	1938	3870000	143	1965	37400000	595	1992	105000000	2795
1912	435000	20	1939	4340000	100	1966	40700000	542	1993	110000000	2463
1913	539000	39	1940	4390000	93	1967	44600000	619	1994	106000000	2447
1914	236000	28	1941	6110000	84	1968	46000000	616	1995	112000000	2331
1915	321000	35	1942	8360000	63	1969	51800000	520	1996	117000000	2266
1916	701000	16	1943	14000000	49	1970	57800000	644	1997	122000000	2218
1917	1030000	19	1944	6960000	75	1971	62100000	692	1998	123000000	2562
1918	818000	15	1945	3430000	80	1972	64900000	872	1999	129000000	2611
1919	569000	37	1946	4360000	95	1973	70400000	783	2000	136000000	2687
1920	901000	45	1947	6320000	125	1974	79600000	786	2001	137000000	2331
1921	318000	47	1948	8360000	143	1975	74800000	966	2002	144000000	2347
1922	701000	43	1949	8230000	159	1976	77400000	1101	2003	153000000	2478
1923	1200000	33	1950	8180000	143	1977	81900000	1113	2004	164000000	2346
1924	1160000	43	1951	10900000	144	1978	81000000	1142	2005	179000000	2219
1925	1380000	81	1952	12800000	220	1979	85500000	1018	2006	199000000	2110
1926	1380000	57	1953	13800000	184	1980	89200000	1513	2007	202000000	2198

¹⁵ Aluminum ore and aluminum oxide [128].

¹⁶ Activity represents world production of bauxite, defined at usgs.gov as "... world mine production of bauxite is reported on a 'dried bauxite equivalents' basis. U.S. bauxite production data are withheld from the total for the years 1989 to the most recent. Alumina world production data is reported as 'quantity produced' (alumina), for the years 1968-71, and as 'calcined alumina equivalents' for the years after 1971. All data are reported in the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹⁷ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Bauxite or alumina were used as keywords found in the patent title or abstract by year of publication.

Table A3.12. Correlation Eq.(A1.1) terms calculated from Table A3.9 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
4.731E+09	83680	5.05E+17	1.6E+08	8.62E+12	2.974E+17	95135138	4.95E+12	0.930961	86.66877

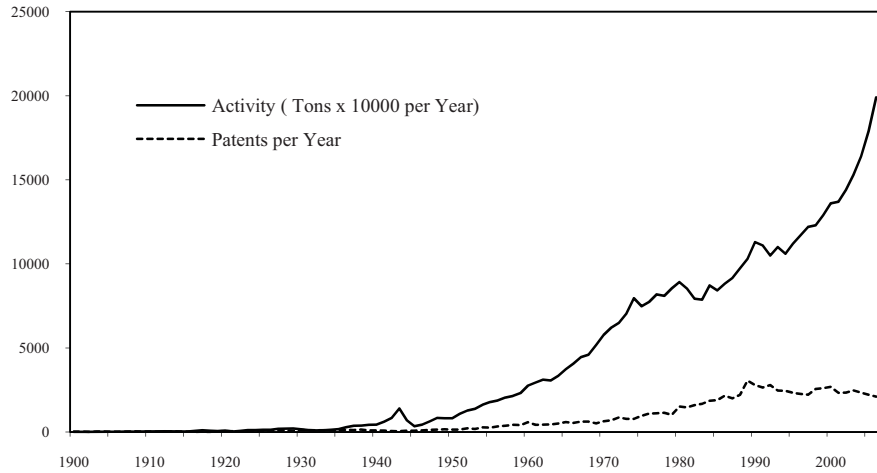


Figure A3.22. Bauxite and Alumina Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

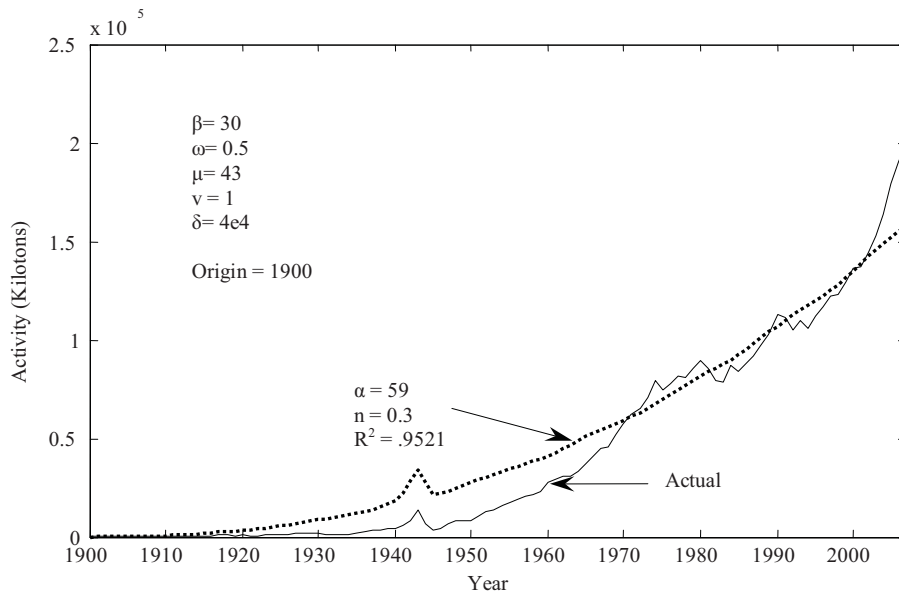


Figure A3.23. USGS World Bauxite/Alumina Production. World bauxite/alumina production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters.

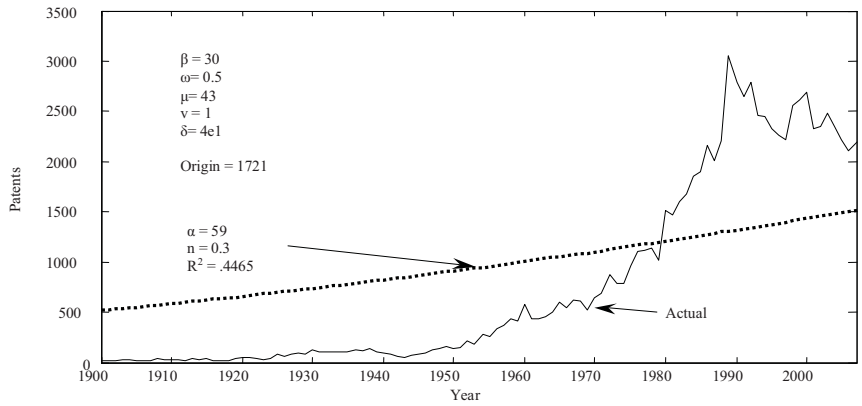


Figure A3.24. EPO Worldwide Patent Search: Bauxite or Alumina in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

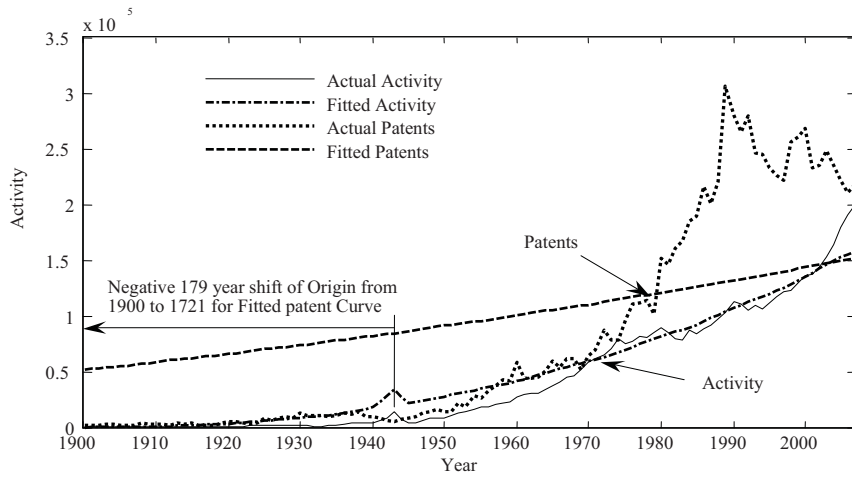


Figure A3.25. Bauxite/Alumina Best-Fit Activity and Patents. Illustrates bauxite/Alumina best-fit origin shift.

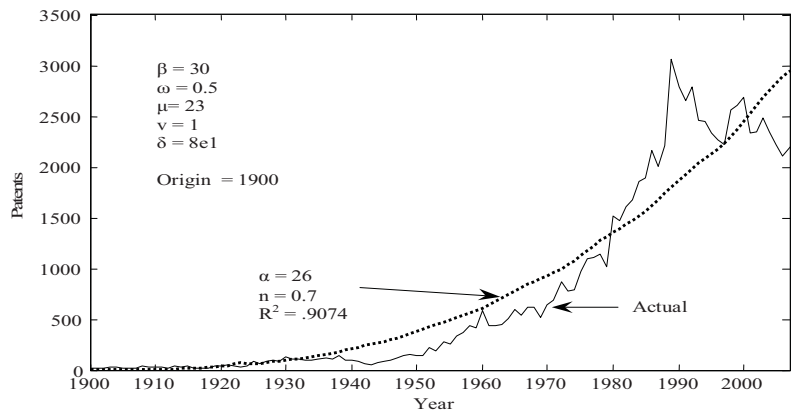


Figure A3.26. Bauxite/Alumina Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.13 Beryllium Activity¹⁸ and Patents¹⁹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	279	54	1981	385	80
1901			1928			1955	323	43	1982	327	115
1902			1929			1956	468	64	1983	366	111
1903			1930			1957	410	74	1984	359	107
1904			1931			1958	279	87	1985	326	125
1905			1932			1959	406	83	1986	356	124
1906			1933			1960	446	117	1987	345	102
1907			1934			1961	468	133	1988	332	114
1908			1935	16	71	1962	399	133	1989	301	112
1909			1936	17	50	1963	265	149	1990	284	103
1910			1937	15	52	1964	178	178	1991	263	107
1911			1938	42	72	1965	222	144	1992	278	115
1912			1939	36	104	1966	165	152	1993	243	117
1913			1940	87	79	1967	197	183	1994	218	137
1914			1941	164	58	1968	263	168	1995	247	118
1915			1942	120	43	1969	322	144	1996	255	101
1916			1943	218	31	1970	249	145	1997	276	114
1917			1944	118	29	1971	210	156	1998	289	160
1918			1945	39	33	1972	157	148	1999	248	133
1919			1946	68	58	1973	144	151	2000	202	163
1920			1947	57	55	1974	126	140	2001	120	169
1921			1948	99	67	1975	119	124	2002	101	161
1922			1949	183	68	1976	93	100	2003	107	170
1923			1950	269	33	1977	103	103	2004	111	128
1924			1951	243	76	1978	105	110	2005	137	116
1925			1952	301	65	1979	96	81	2006	179	97
1926			1953	298	44	1980	373	97	2007	179	123

Table A3.14. Correlation Eq.(A1.1) terms calculated from Table A3.13 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
16089	7671	4540781	923695	1729386	994809.48	117609.5	38718.62	0.113195	1.281319

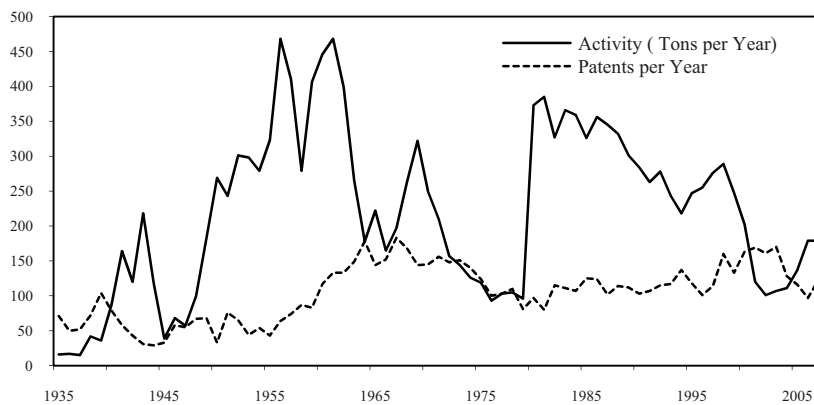


Figure A3.27. Beryllium Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

¹⁸ Activity represents world production of beryllium, defined at usgs.gov as "...the estimated beryllium content of beryllium-bearing ores produced throughout the world. World mine production data are based on a beryllium metal equivalent of 4 percent Be in beryl and bertrandite ores, reported as equivalent to beryl ore containing 11 percent BeO. Data are not available prior to the year 1935. U.S. production data for the years 1964-67 and 1969-79 are not available and are not included in the total. Data are from the MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹⁹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Beryllium was used as the keyword found in the patent title or abstract by year of publication.

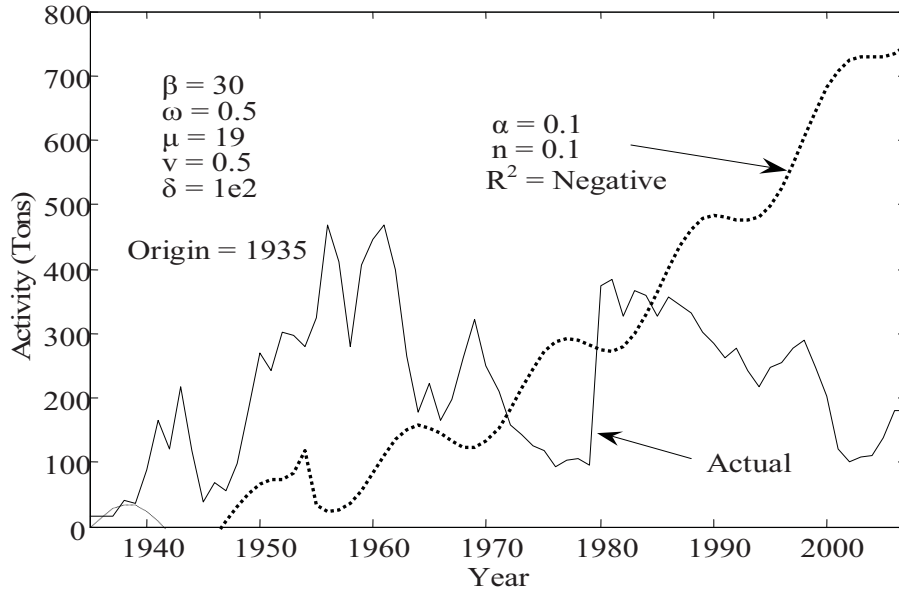


Figure A3.28. USGS World Beryllium Production. World beryllium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. R^2 has a negative value indicating Stage IV. No best-fit for the patent data was obtainable.

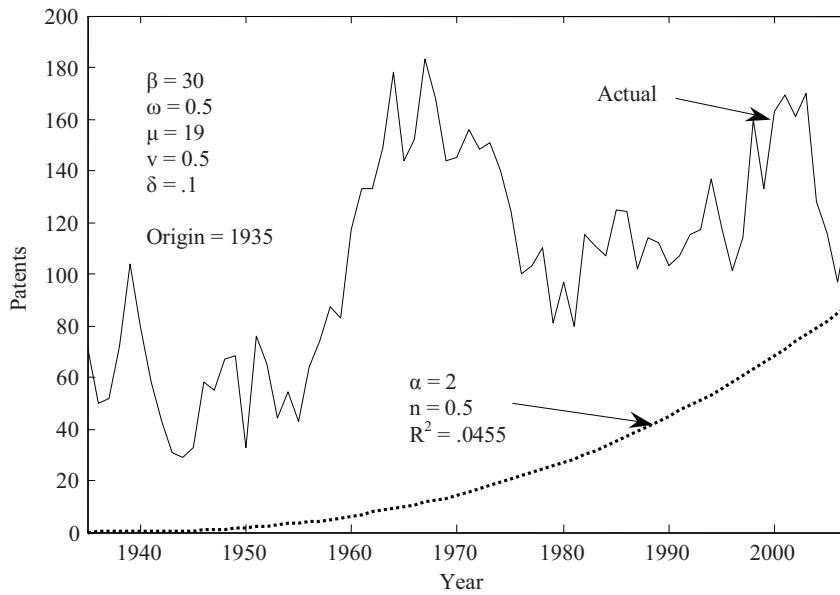


Figure A3.29. Beryllium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.15 Bismuth Activity²⁰ and Patents²¹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	1680	57	1981	3750	236
1901			1928			1955	1910	43	1982	4110	215
1902			1929			1956	2400	60	1983	3980	259
1903			1930			1957	2270	66	1984	3480	240
1904			1931			1958	2090	61	1985	4410	207
1905			1932			1959	2270	77	1986	3660	234
1906			1933			1960	2400	124	1987	3170	221
1907			1934			1961	2590	107	1988	3220	261
1908			1935			1962	3040	110	1989	3650	339
1909			1936			1963	2530	110	1990	3440	497
1910			1937	700	46	1964	2890	149	1991	3230	385
1911			1938	1000	51	1965	2960	134	1992	2870	506
1912			1939	1300	47	1966	3110	118	1993	3550	472
1913			1940	1400	39	1967	3380	124	1994	3410	430
1914			1941	1400	43	1968	3770	122	1995	3430	387
1915			1942	1700	15	1969	3760	94	1996	3600	457
1916			1943	1400	14	1970	3720	103	1997	4490	468
1917			1944	1200	18	1971	3830	129	1998	3990	504
1918			1945	1100	32	1972	4000	132	1999	5490	545
1919			1946	940	22	1973	3720	132	2000	3760	542
1920			1947	1500	26	1974	4820	135	2001	4420	570
1921			1948	1500	45	1975	3980	153	2002	4600	603
1922			1949	1500	35	1976	3940	145	2003	5100	651
1923			1950	1400	31	1977	4480	157	2004	5600	665
1924			1951	1770	32	1978	4250	155	2005	5400	602
1925			1952	1770	37	1979	3420	162	2006	5700	588
1926			1953	2090	34	1980	3610	221	2007	6300	671

Table A3.16. Correlation Eq.(A1.1) terms calculated from Table A3.15 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
222300	15502	8.17E+08	6186822	62618120	120900631	2802146	14081576	0.765053	58.53064

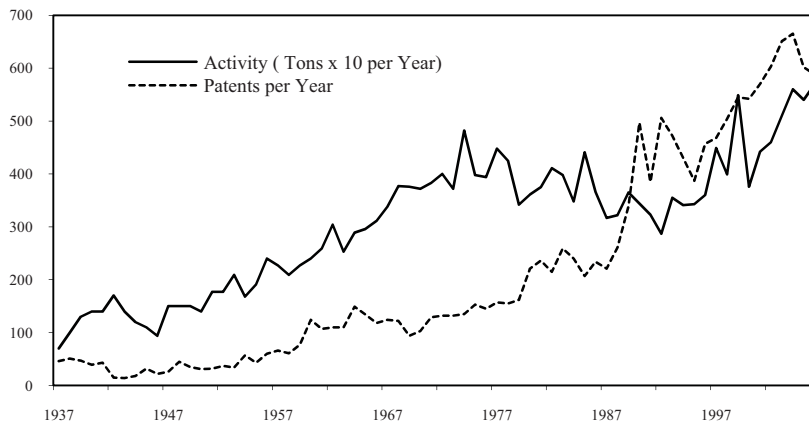


Figure A3.30. Bismuth Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

²⁰ Activity represents world production of bismuth, defined at usgs.gov as "...bismuth content of world mine production. Data were not available prior to 1912 or for the years 1922-36. Data for the years 1912-21 and 1972-2003 exclude U.S. production. Data are from the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]. Data for 2004 is an unpublished revision published by the Commodity Specialist." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

²¹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Bismuth was used as the keyword found in the patent title or abstract by year of publication.

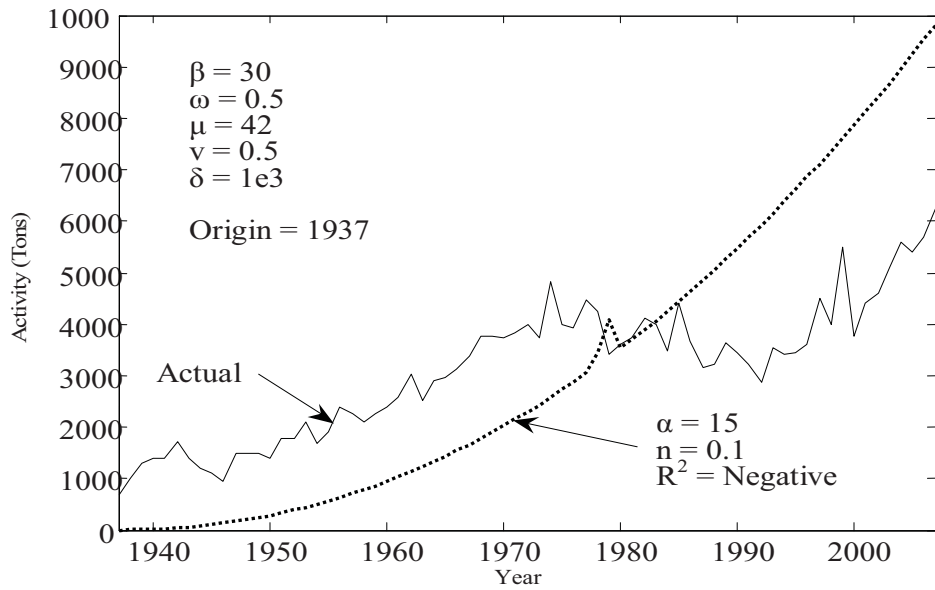


Figure A3.31. USGS World Bismuth Production. World bismuth production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. R^2 has a negative value indicating possible Stage IV. No best-fit for the patent data was obtainable.

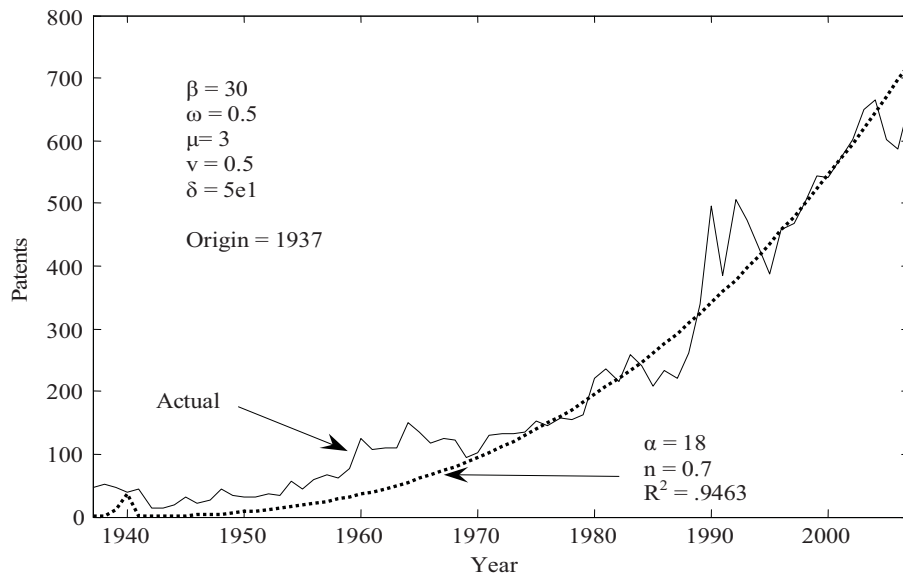


Figure A3.32. Bismuth Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.17 Boron Activity²² and Patents²³

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981	2560000	796
1901			1928			1955			1982	2270000	946
1902			1929			1956			1983	2240000	1126
1903			1930			1957			1984	2510000	1147
1904			1931			1958			1985	2510000	1425
1905			1932			1959			1986	2510000	1383
1906			1933			1960			1987	2690000	1536
1907			1934			1961			1988	2990000	1705
1908			1935			1962			1989	2990000	1899
1909			1936			1963			1990	2910000	1759
1910			1937			1964	172000	412	1991	2960000	1785
1911			1938			1965	189000	466	1992	2670000	2097
1912			1939			1966	209000	435	1993	2640000	1691
1913			1940			1967	221000	437	1994	3810000	1737
1914			1941			1968	232000	378	1995	4020000	1543
1915			1942			1969	251000	359	1996	4330000	1558
1916			1943			1970	257000	400	1997	4580000	1537
1917			1944			1971	284000	410	1998	4570000	1764
1918			1945			1972	314000	489	1999	4460000	1902
1919			1946			1973	342000	422	2000	4600000	2147
1920			1947			1974	328000	385	2001	4740000	2182
1921			1948			1975	354000	473	2002	4580000	2319
1922			1949			1976	2340000	505	2003	4750000	2342
1923			1950			1977	2730000	498	2004	4960000	2140
1924			1951			1978	2660000	549	2005	4840000	1990
1925			1952			1979	2520000	553	2006	3580000	1926
1926			1953			1980	2610000	782	2007	3840000	2089

Table A3.18. Correlation Eq.(A1.1) terms calculated from Table A3.17 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
112123000	54424	4E+14	88106536	1.81E+11	1.143E+14	20788996	4.24E+10	0.869138	75.54004

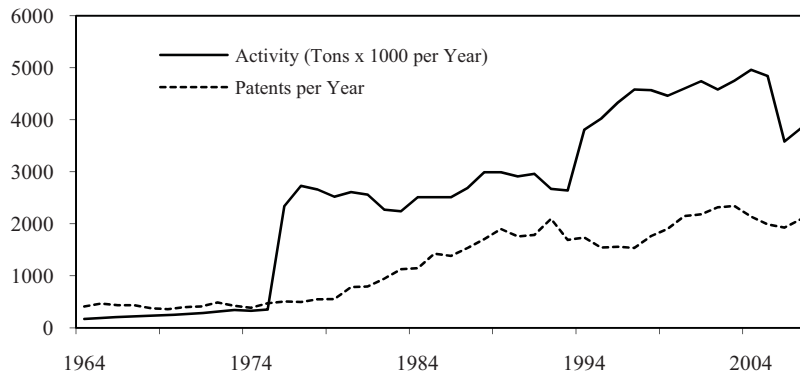


Figure A3.33. Boron Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

²² Activity represents world production of boron, defined at usgs.gov as "...world mine production. For most years, world mine production data are reported in gross weight. Data could not be converted to contained B₂O₃ because various boron units are used when reporting the minerals and compounds of boron. World production data are not reported for the years 1914-64. Data reported in the MR [*Minerals Resources of the United States*] and MYB [*Minerals Yearbook*] cover the years 1900-13 and 1976 to the most recent and are all gross weight data. Data for the years 1964-75 are calculated B₂O₃ content and are reported in the 1975 and 1980 MFP [*Mineral Facts and Problems*]. World production data from 2006 to most recent does not include U.S. data." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

²³ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Boron was used as the keyword found in the patent title or abstract by year of publication.

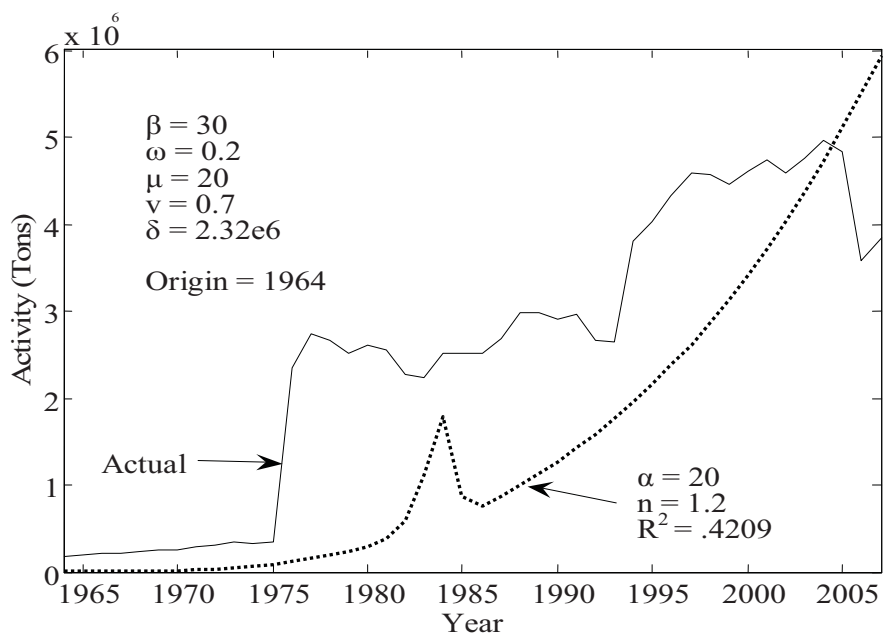


Figure A3.34. USGS World Production. World boron production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

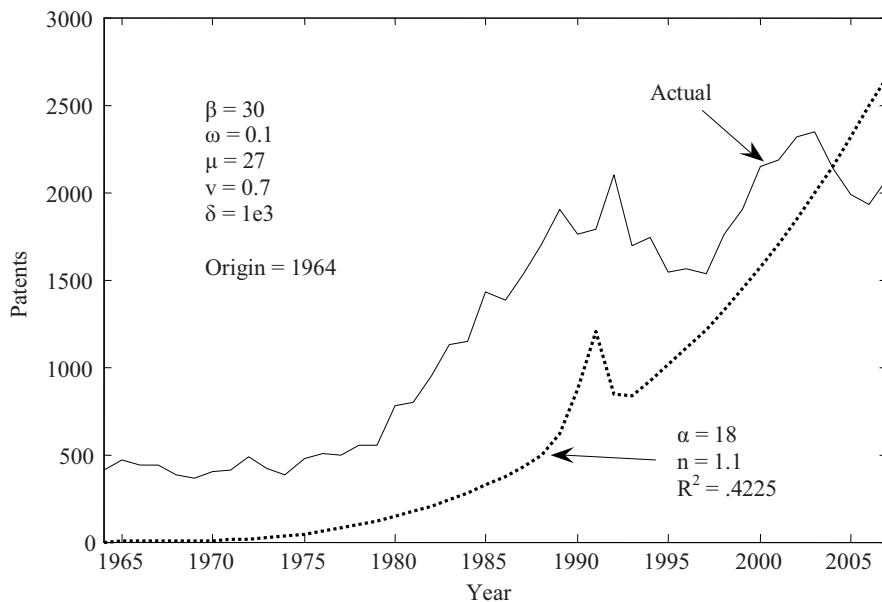


Figure A3.35. Boron Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.19 Cadmium Activity²⁴ and Patents²⁵

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	14	5	1927	661	39	1954	7300	185	1981	17400	672
1901	13	8	1928	1500	65	1955	8390	211	1982	16400	760
1902	13	6	1929	2180	70	1956	9070	242	1983	17600	809
1903	17	10	1930	2480	80	1957	9390	275	1984	19600	892
1904	25	6	1931	1230	104	1958	9800	241	1985	19100	941
1905	25	6	1932	1040	87	1959	10200	252	1986	19100	1039
1906	21	11	1933	1950	115	1960	11100	399	1987	19000	1070
1907	39	17	1934	2480	124	1961	11700	326	1988	21900	1171
1908	37	12	1935	3150	136	1962	11700	318	1989	21400	1316
1909	39	16	1936	3390	146	1963	11800	351	1990	20200	1223
1910	43	10	1937	3970	127	1964	12700	418	1991	20900	1241
1911	55	7	1938	4010	155	1965	11900	450	1992	19600	1516
1912	67	14	1939	4580	140	1966	13000	428	1993	18700	1565
1913	62	7	1940	5220	139	1967	13200	499	1994	18200	1615
1914	80	8	1941	5220	92	1968	15000	458	1995	20100	1829
1915	78	11	1942	5030	44	1969	17600	429	1996	18900	2010
1916	119	3	1943	5380	48	1970	16500	476	1997	20300	2132
1917	172	7	1944	5320	45	1971	15400	468	1998	20200	2491
1918	165	13	1945	4760	75	1972	16700	502	1999	20000	2439
1919	89	11	1946	4050	93	1973	17200	467	2000	20300	3090
1920	81	24	1947	4930	85	1974	17300	408	2001	20000	3045
1921	54	20	1948	4870	127	1975	15200	423	2002	17800	3566
1922	128	28	1949	5220	148	1976	17000	458	2003	18400	3272
1923	248	27	1950	6010	115	1977	18300	474	2004	18700	3225
1924	245	26	1951	6070	130	1978	17300	543	2005	20200	2898
1925	4116	46	1952	6170	163	1979	18700	397	2006	19900	2743
1926	960	33	1953	7060	135	1980	18200	633	2007	20400	2295

Table A3.20. Correlation Eq.(A1.1) terms calculated from Table A3.19 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
1010886	64815	1.62E+10	1.22E+08	1.16E+09	6.721E+09	83019689	5.53E+08	0.740132	54.77951

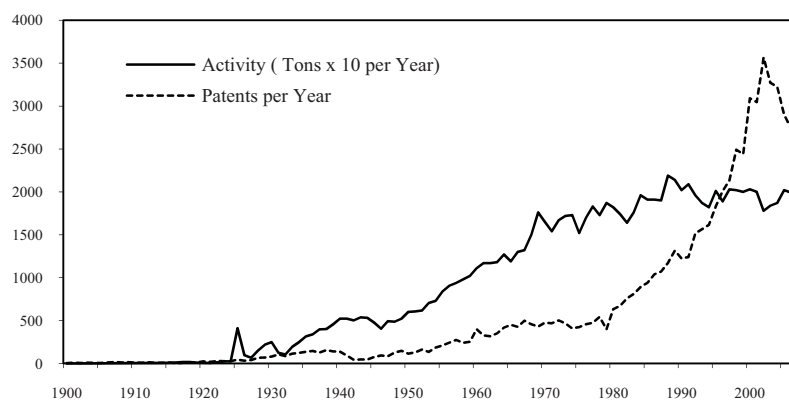


Figure A3.36. Cadmium Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

²⁴ Activity represents world production of cadmium, defined at usgs.gov as "... world refinery production of cadmium. World production begins in 1932. Data prior to 1932 are production data from selected countries. Data are from the MR [*Minerals Resources of the United States*] and MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

²⁵ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Cadmium or Cd were used as keywords found in the patent title or abstract by year of publication.

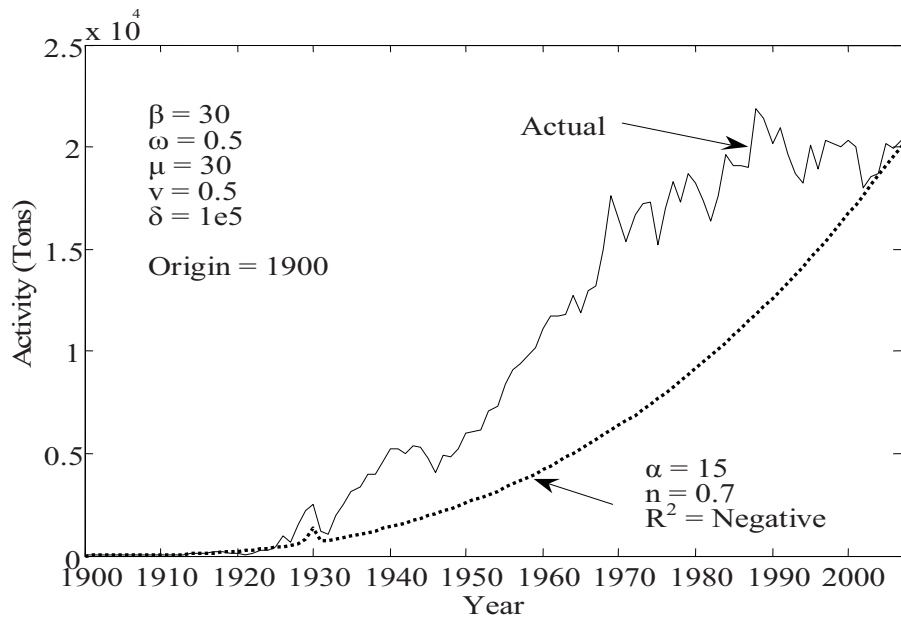


Figure A3.37. USGS World Cadmium Production. World cadmium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. R^2 has a negative value indicating possible Stage IV. No best-fit for the patent data was obtainable.

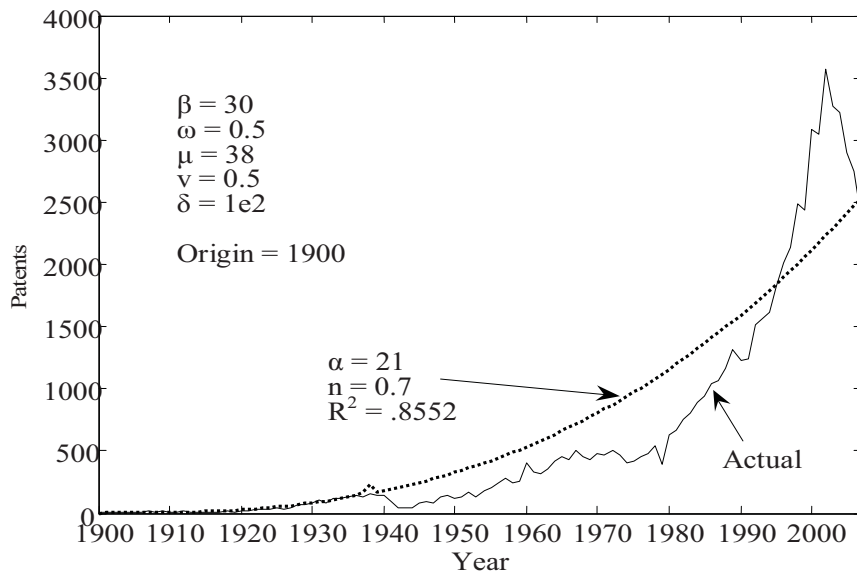


Figure A3.38. Cadmium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.21. Chromium Activity²⁶ and Patents²⁷

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	16500	42	1927	124000	222	1954	924000	413	1981	2550000	2850
1901	27900	36	1928	140000	289	1955	1040000	479	1982	2390000	3420
1902	26400	44	1929	197000	326	1956	1200000	526	1983	2540000	3693
1903	29500	42	1930	173000	391	1957	1370000	601	1984	2950000	3649
1904	36600	48	1931	127000	458	1958	1130000	578	1985	3180000	4006
1905	44500	57	1932	101000	413	1959	1150000	582	1986	3530000	4428
1906	49700	45	1933	123000	342	1960	1250000	836	1987	3450000	4546
1907	34700	55	1934	183000	385	1961	1220000	739	1988	3870000	4563
1908	20700	48	1935	241000	441	1962	1280000	732	1989	4320000	5068
1909	33300	42	1936	317000	359	1963	1170000	777	1990	3950000	4750
1910	33600	61	1937	392000	415	1964	1290000	833	1991	4060000	4776
1911	25100	67	1938	362000	439	1965	1490000	968	1992	3420000	5239
1912	38000	65	1939	347000	361	1966	1360000	863	1993	3080000	5130
1913	45500	62	1940	457000	270	1967	1430000	1083	1994	3090000	5336
1914	48500	49	1941	509000	247	1968	1560000	1042	1995	4530000	5225
1915	57400	53	1942	637000	207	1969	1670000	1003	1996	3660000	5028
1916	87000	39	1943	542000	177	1970	1910000	1176	1997	4330000	4872
1917	81300	52	1944	411000	173	1971	2000000	1309	1998	4460000	5380
1918	96500	48	1945	318000	190	1972	1970000	1462	1999	4810000	5286
1919	52900	62	1946	352000	214	1973	2030000	1341	2000	4750000	5708
1920	53200	76	1947	521000	246	1974	2230000	1242	2001	3740000	5690
1921	41400	122	1948	644000	331	1975	2530000	1383	2002	4510000	6275
1922	43300	153	1949	650000	332	1976	2430000	1530	2003	4770000	5887
1923	63500	178	1950	720000	239	1977	2600000	1759	2004	5480000	5989
1924	90200	129	1951	823000	290	1978	2990000	1916	2005	5810000	5331
1925	95300	160	1952	963000	430	1979	2590000	1976	2006	5850000	5383
1926	112000	174	1953	1130000	327	1980	2830000	2836	2007	6620000	5703

Table A3.22 Correlation Eq.(A1.1) terms calculated from Table A3.21 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
169232500	173719	5.73E+14	7.2E+08	6.21E+11	3.074E+14	4.4E+08	3.49E+11	0.949454	90.14634

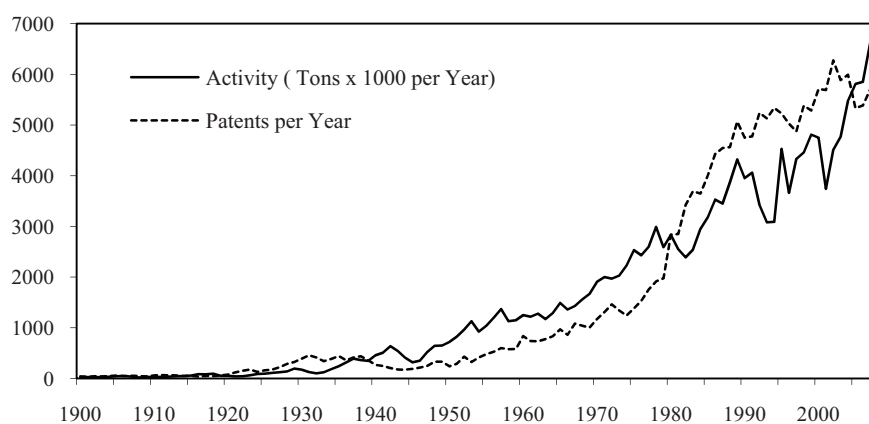


Figure A3.39. Chromium Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

²⁶ Activity represents world production of chromium, defined at usgs.gov as “...an estimate of world chromite ore mine production measured in contained chromium. World production reported in gross weight was converted to contained chromium by assuming that its chromic oxide content was the same as that of chromite ore imported into the United States. Before content of chromite ore was reported, a time-averaged value was used.” Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

²⁷ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Chromium, Cr or Chrome were used as keywords found in the patent title or abstract by year of publication.

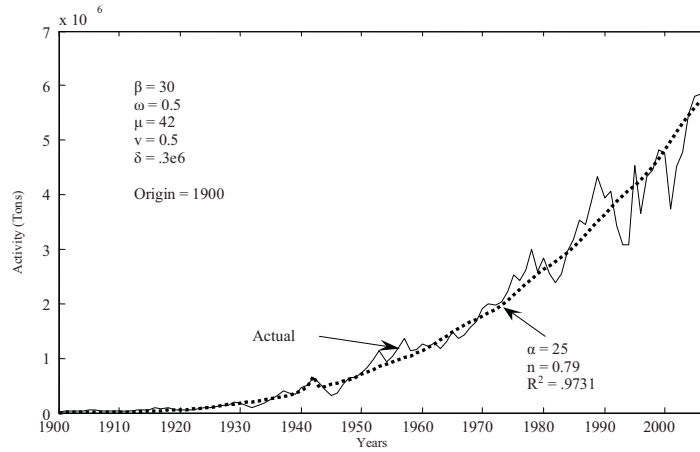


Figure A3.40. USGS World Chromium Production. World chromium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

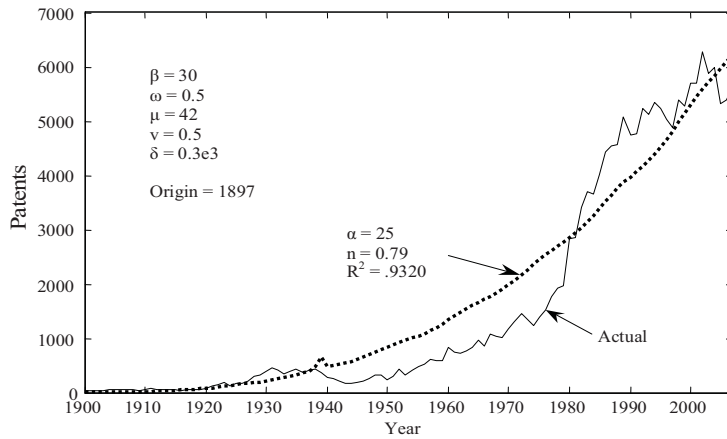


Figure A3.41. EPO Worldwide Patent Search: Chromium , Cr or Chrome in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

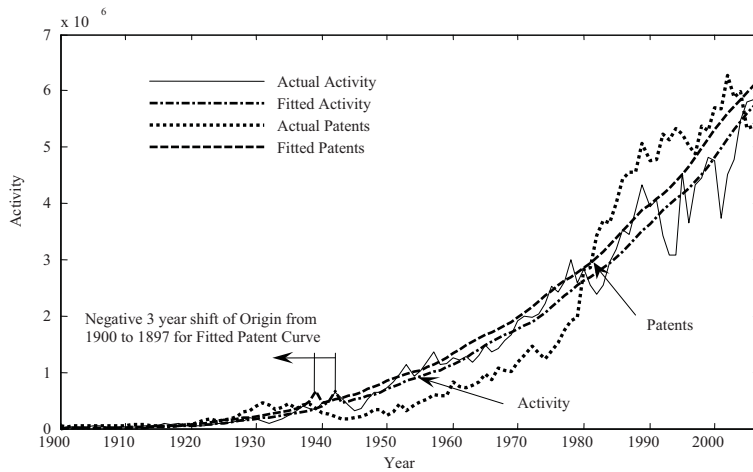


Figure A3.42 Chromium Best-Fit Activity and Patents. Illustrates best-fit origin shift.

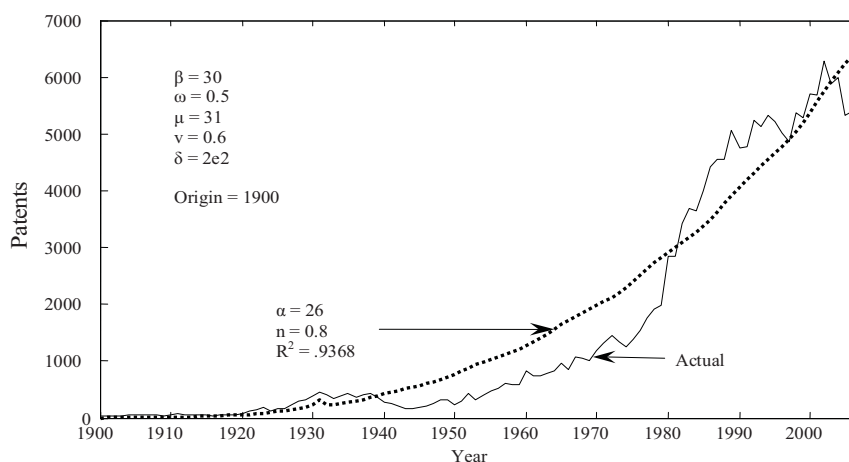


Figure A3.43. Chromium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.23 Cobalt Activity²⁸ and Patents²⁹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	1180	72	1954	13100	298	1981	30700	1018
1901	180	11	1928	1180	99	1955	13300	319	1982	24600	1044
1902	540	9	1929	1360	127	1956	14400	388	1983	37900	1094
1903	640	8	1930	1270	151	1957	14400	431	1984	40900	1086
1904	540	9	1931	910	205	1958	12600	389	1985	47400	1109
1905	450	8	1932	1090	180	1959	14800	401	1986	50200	1098
1906	450	15	1933	1270	173	1960	14200	511	1987	41200	1117
1907	910	11	1934	1450	169	1961	14400	449	1988	43800	1099
1908	1360	15	1935	2000	182	1962	17100	457	1989	42900	1169
1909	1450	8	1936	2720	151	1963	14500	417	1990	42300	1117
1910	1000	19	1937	3800	191	1964	17800	467	1991	33300	1092
1911	820	13	1938	4500	199	1965	19000	526	1992	28000	1234
1912	860	7	1939	4500	187	1966	21800	441	1993	21900	1216
1913	820	16	1940	5000	120	1967	20500	469	1994	18000	1106
1914	360	19	1941	4000	129	1968	19600	457	1995	24500	1144
1915	230	22	1942	3500	66	1969	20200	396	1996	26200	1186
1916	410	12	1943	4200	60	1970	24200	415	1997	27400	1338
1917	360	16	1944	3900	60	1971	25100	440	1998	35300	1655
1918	450	16	1945	4700	90	1972	24800	610	1999	32700	1637
1919	360	24	1946	3500	143	1973	29400	620	2000	37900	1899
1920	360	25	1947	5000	129	1974	30900	563	2001	47900	1892
1921	180	33	1948	6100	181	1975	30800	727	2002	50700	1998
1922	820	29	1949	5900	181	1976	21400	726	2003	52900	2011
1923	640	37	1950	7170	179	1977	21500	765	2004	58600	1853
1924	1090	38	1951	8440	183	1978	26800	913	2005	63400	1716
1925	1090	56	1952	10100	282	1979	29900	859	2006	65900	1634
1926	820	41	1953	11300	189	1980	31300	981	2007	65500	1749

²⁸ Activity represents world mine production of cobalt, defined at usgs.gov as "...cobalt content of refined products or the cobalt content, recoverable cobalt content, or recovered cobalt content of mined ores, concentrates, or intermediate products depending on the producing country and year.....Data for the years 1901-36 are from IC [U.S. Bureau of Mines Information Circular] 8103. Data for the years 1937 to the most recent are from the MYB [Minerals Yearbook]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

²⁹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Cobalt was used as the keyword found in the patent title or abstract by year of publication.

Table A3.24. Correlation Eq.(A1.1) terms calculated from Table A3.23 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
1777030	56341	6.21E+10	63953703	1.92E+09	3.258E+10	34287270	9.8E+08	0.926925	85.91898

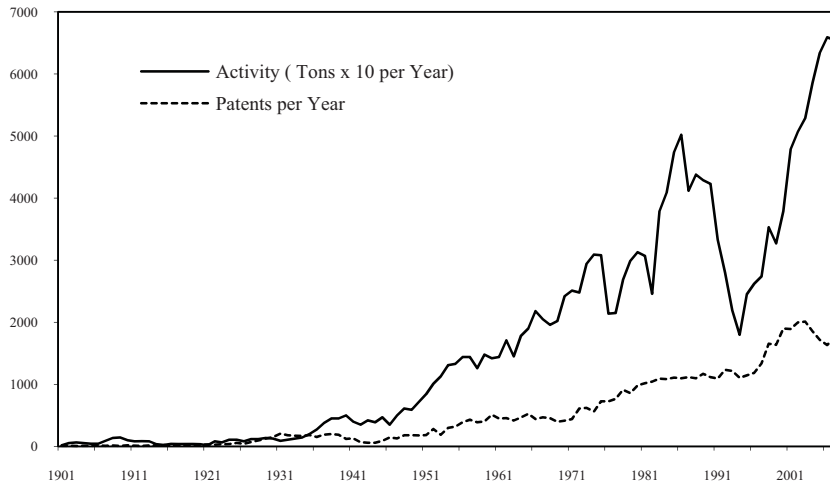


Figure A3.44 Cobalt Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

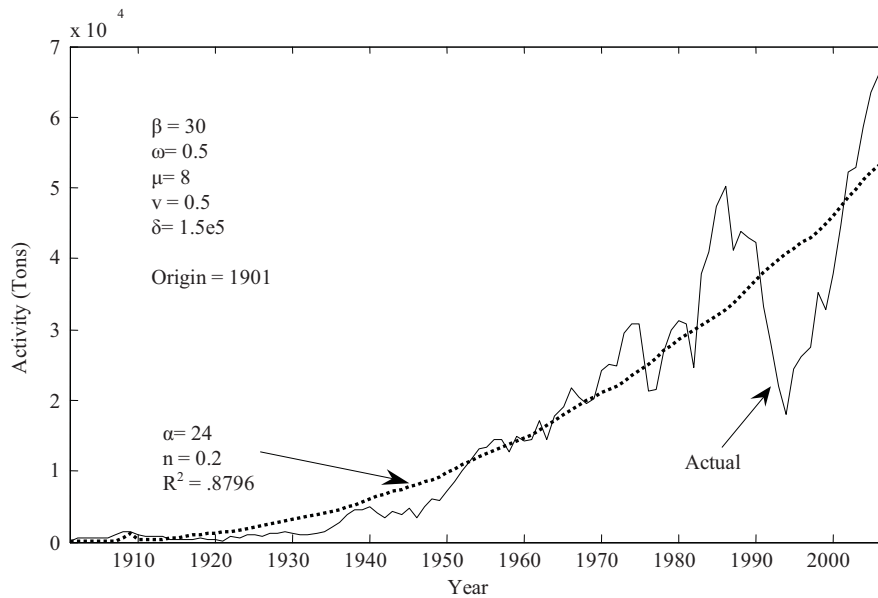


Figure A3.45. USGS World Cobalt Production. World cobalt production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

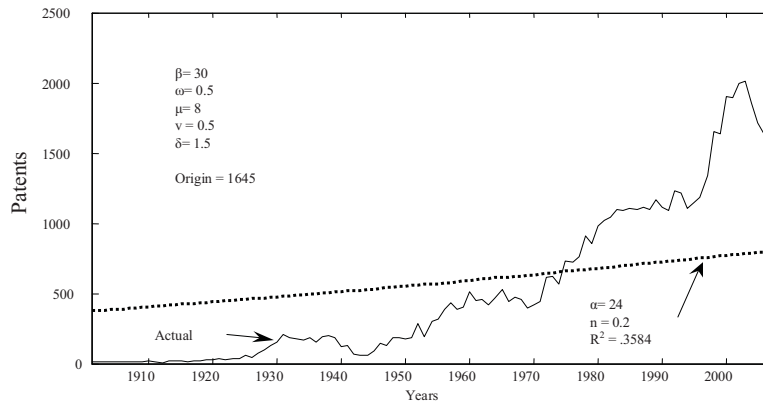


Figure A3.46. EPO Worldwide Patent Search: Cobalt in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

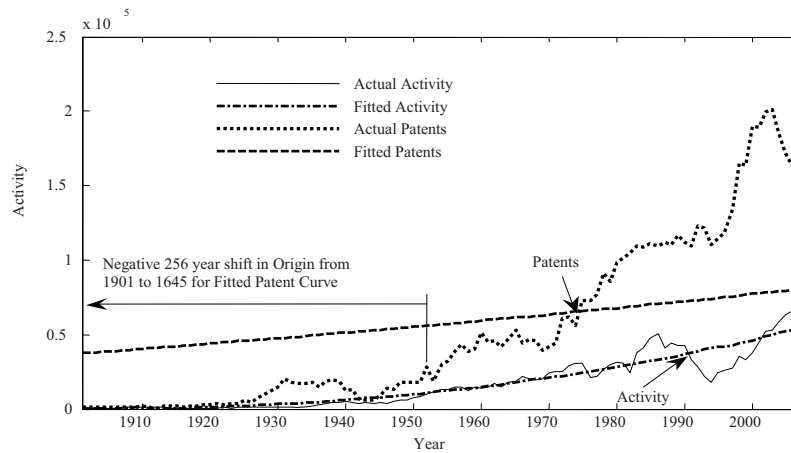


Figure A3.47 Cobalt Best-Fit Activity and Patents. Illustrates best-fit origin shift.

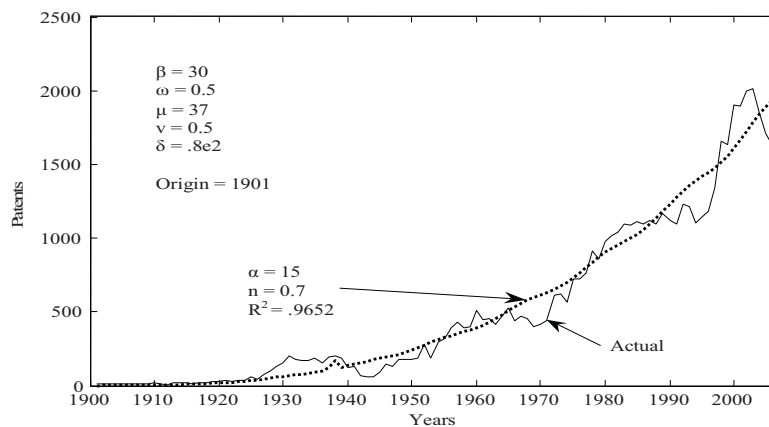


Figure A3.48. Cobalt Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.25. Copper Activity³⁰ and Patents³¹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	495000	165	1927	1520000	447	1954	2640000	773	1981	7690000	4182
1901	526000	169	1928	1730000	511	1955	2900000	824	1982	7580000	4673
1902	555000	186	1929	1950000	538	1956	3200000	964	1983	7610000	4684
1903	596000	232	1930	1610000	640	1957	3300000	1016	1984	7810000	5125
1904	660000	221	1931	1400000	773	1958	3190000	896	1985	7990000	5574
1905	713000	236	1932	909000	646	1959	3430000	925	1986	7940000	6094
1906	724000	198	1933	1050000	611	1960	3940000	1301	1987	8240000	5766
1907	721000	254	1934	1280000	601	1961	4090000	1190	1988	8720000	7025
1908	744000	257	1935	1500000	695	1962	4220000	1100	1989	9040000	9723
1909	828000	219	1936	1720000	675	1963	4290000	1255	1990	9200000	8660
1910	858000	200	1937	2290000	670	1964	4450000	1335	1991	9330000	8422
1911	890000	194	1938	1990000	749	1965	4660000	1548	1992	9470000	9392
1912	1000000	218	1939	2130000	637	1966	4580000	1424	1993	9490000	8266
1913	996000	241	1940	2400000	560	1967	4630000	1806	1994	9500000	8197
1914	938000	197	1941	2480000	430	1968	5010000	1751	1995	10000000	7604
1915	1060000	182	1942	2590000	322	1969	5520000	1677	1996	11000000	7831
1916	1420000	125	1943	2620000	316	1970	5900000	1992	1997	11500000	7423
1917	1430000	117	1944	2460000	300	1971	5940000	2066	1998	12100000	8551
1918	1430000	120	1945	2110000	368	1972	6540000	2404	1999	12800000	8868
1919	994000	228	1946	1780000	451	1973	6920000	2240	2000	13200000	10521
1920	959000	234	1947	2130000	490	1974	7100000	2049	2001	13700000	10566
1921	558000	310	1948	2210000	673	1975	6740000	2235	2002	13600000	11246
1922	884000	260	1949	2140000	649	1976	7260000	2526	2003	13800000	11222
1923	1270000	310	1950	2380000	495	1977	7420000	2820	2004	14700000	11303
1924	1360000	327	1951	2490000	576	1978	7280000	2840	2005	15000000	10065
1925	1530000	386	1952	2570000	731	1979	7350000	2710	2006	15100000	9776
1926	1510000	368	1953	2600000	564	1980	7200000	3820	2007	15400000	10130

Table A3.26. Correlation Eq.(A1.1) terms calculated from Table A3.25 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
510898000	289648	4.27E+15	2.04E+09	2.82E+12	1.851E+15	1.26E+09	1.45E+12	0.950672	90.37776

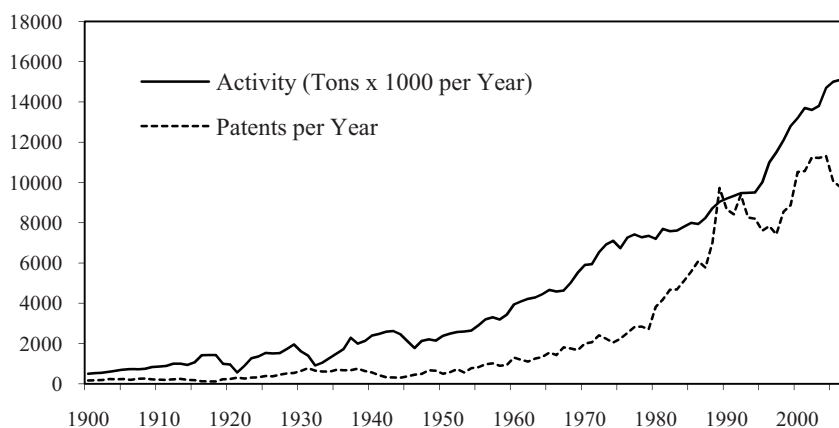


Figure A3.49 Copper Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

³⁰ Activity represents world production of copper, defined at usgs.gov as follows. “World mine production is based on a compilation of available data published in the MR and the MYB and generally reflects the copper content of concentrates, precipitates, and electrowon copper. For some countries, including the United States, recoverable copper content is used. For other countries, such as Chile, data includes copper content of non-duplicative mine and metal products produced from domestic ores and concentrates.” (usgs.gov) Data is in metric tons, as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

³¹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Copper or Cu were used as keywords found in the patent title or abstract by year of publication.

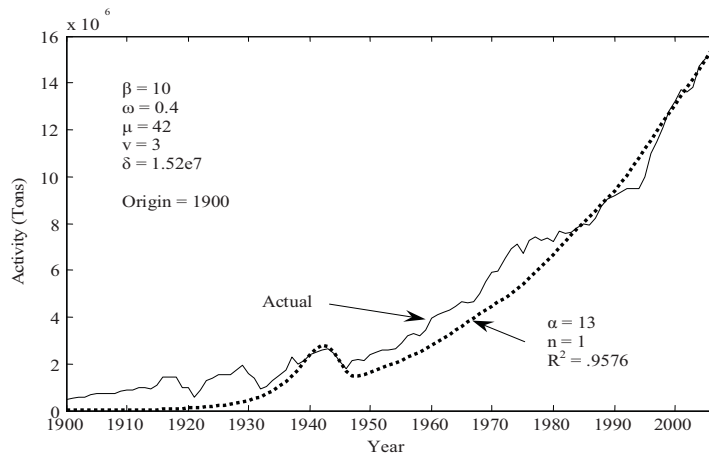


Figure A3.50. USGS World Copper Production. World copper production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

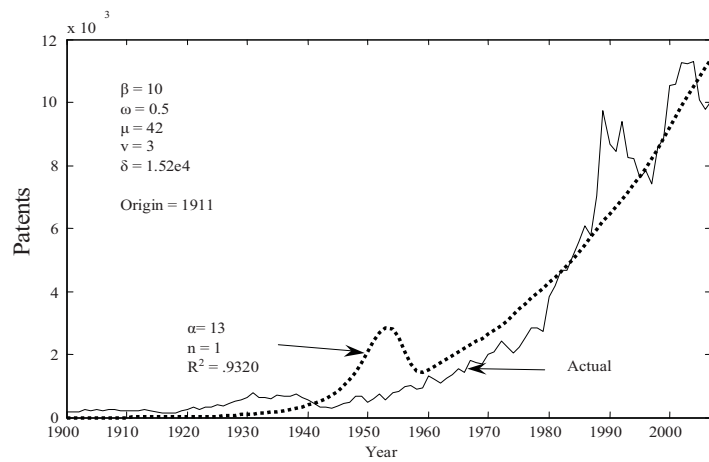


Figure A3.51. EPO Worldwide Patent Search: Copper or Cu in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

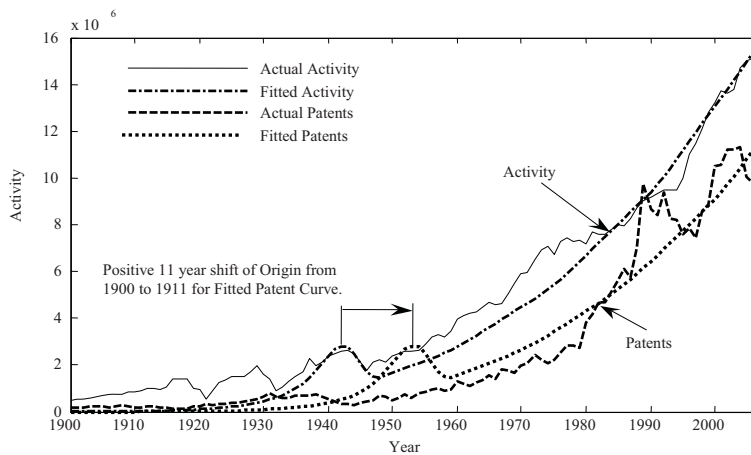


Figure A3.52. Copper Best-Fit Activity and Patents. Illustrates best-fit origin shift.

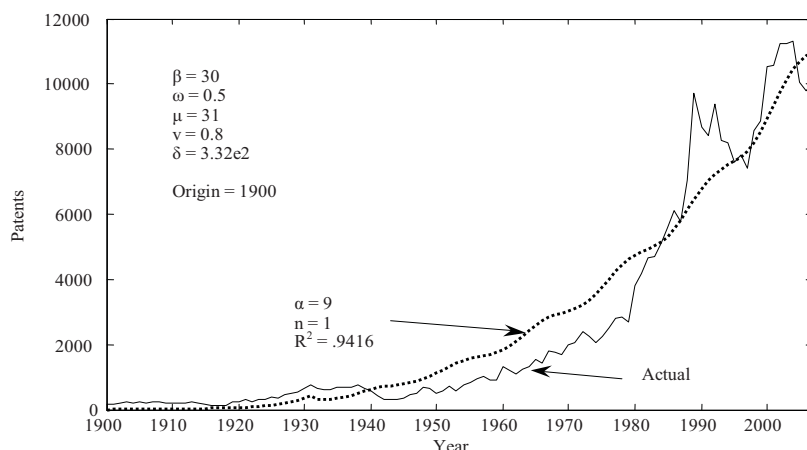


Figure A3.53. Copper Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.27 Feldspar³² Activity³³ and Patents³⁴

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	340000	0	1954	945000	5	1981	3230000	18
1901			1928	366000	4	1955	1070000	7	1982	3520000	18
1902			1929	331000	3	1956	1130000	3	1983	3590000	16
1903			1930	284000	5	1957	1260000	5	1984	3790000	22
1904			1931	239000	2	1958	1210000	9	1985	4030000	23
1905			1932	189000	3	1959	1370000	6	1986	4120000	30
1906			1933	258000	3	1960	1570000	6	1987	4380000	14
1907			1934	280000	2	1961	1630000	11	1988	4840000	31
1908	150000	4	1935	316000	6	1962	1630000	6	1989	5180000	55
1909	168000	3	1936	381000	7	1963	1710000	8	1990	5990000	44
1910	179000	2	1937	426000	8	1964	1890000	5	1991	5670000	53
1911	210000	0	1938	320000	5	1965	2010000	9	1992	5990000	56
1912	206000	1	1939	390000	4	1966	2150000	10	1993	6170000	52
1913	156000	4	1940	400000	2	1967	2040000	18	1994	6490000	62
1914	193000	3	1941	475000	13	1968	2240000	20	1995	7910000	48
1915	155000	3	1942	440000	5	1969	2450000	10	1996	8290000	60
1916	168000	3	1943	440000	2	1970	2530000	17	1997	8650000	64
1917	174000	6	1944	465000	2	1971	2550000	22	1998	9330000	62
1918	136000	7	1945	500000	1	1972	2720000	12	1999	9980000	79
1919	108000	8	1946	675000	5	1973	2770000	21	2000	9540000	103
1920	207000	10	1947	700000	10	1974	3010000	16	2001	11800000	99
1921	167000	9	1948	770000	4	1975	2630000	13	2002	14100000	99
1922	208000	12	1949	660000	8	1976	2800000	20	2003	13600000	124
1923	250000	8	1950	721000	9	1977	2940000	13	2004	15100000	119
1924	362000	3	1951	793000	3	1978	3030000	16	2005	16200000	83
1925	314000	6	1952	803000	9	1979	3110000	14	2006	17600000	80
1926	345000	3	1953	793000	4	1980	3200000	17	2007	18100000	111

³² Silicate minerals linked with potassium, sodium and calcium [128].

³³ Activity represents world production of feldspar, defined at usgs.gov as "...the quantity of feldspar that was produced annually throughout the world as reported in the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]. World production does not include production data for nepheline syenite" Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

³⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Feldspar was used as the keyword found in the patent title or abstract by year of publication.

Table A3.28. Correlation Eq.(A1.1) terms calculated from Table A3.27 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
301396000	2168	2.62E+15	133006	1.8E+10	1.707E+15	86003.76	1.15E+10	0.949019	90.06377

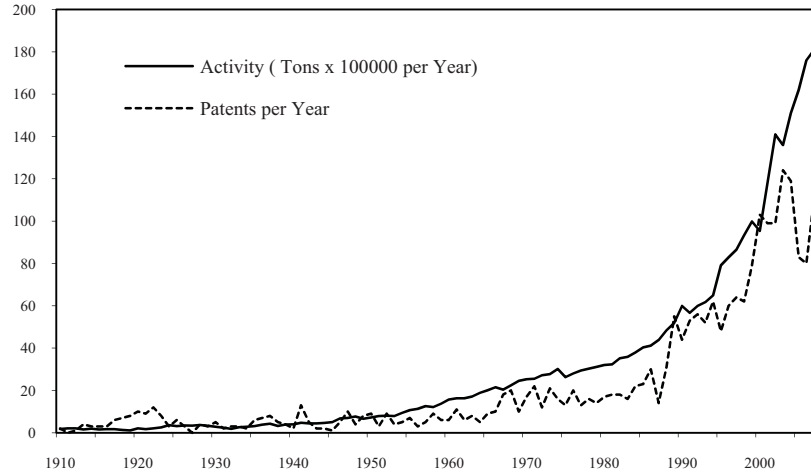


Figure A3.54. Feldspar Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

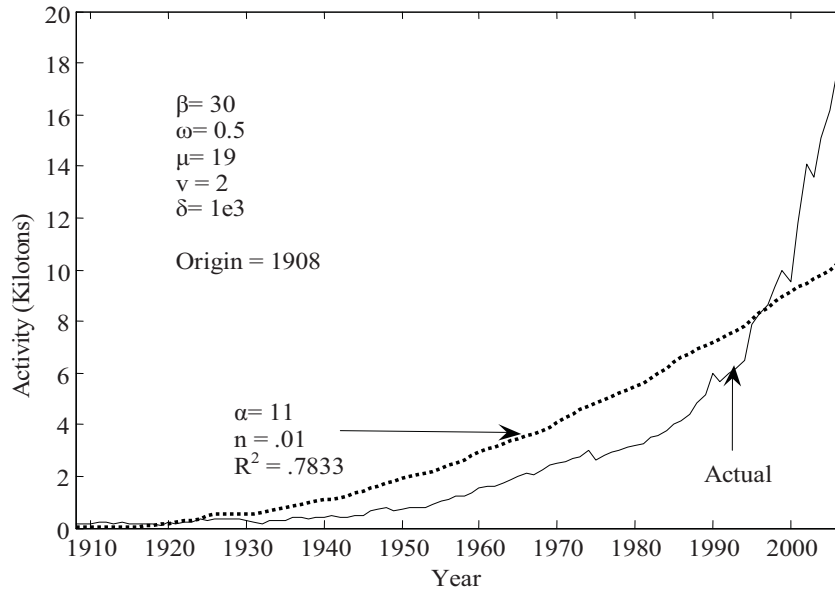


Figure A3.55. USGS World Feldspar Production. World feldspar production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters.

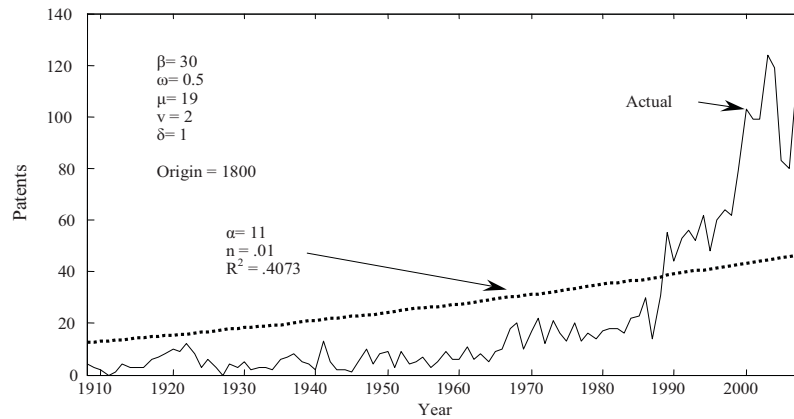


Figure A3.56. EPO Worldwide Patent Search: Feldspar in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

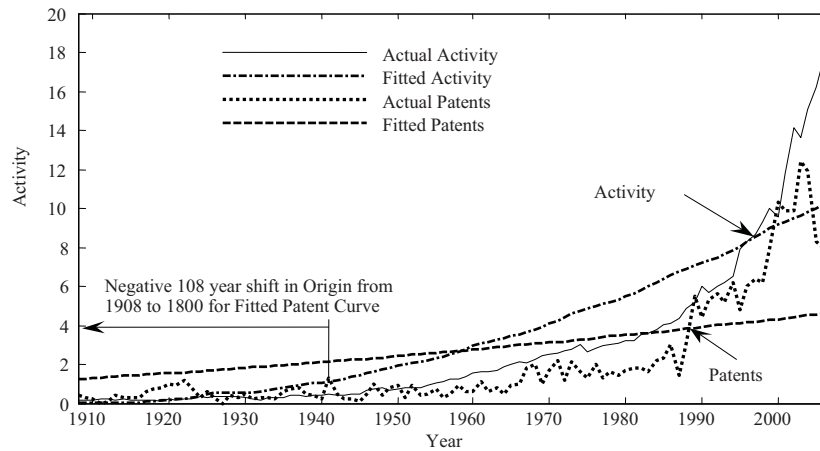


Figure A3.57. Feldspar Best-Fit Activity and Patents. Illustrates best-fit origin shift.

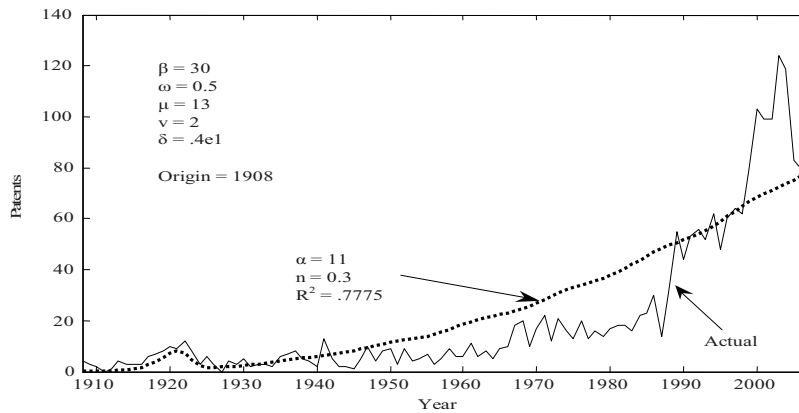


Figure A3.58. Feldspar Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.29 Fluorspar³⁵ Activity³⁶ and Patents³⁷

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	317000	6	1954	1220000	7	1981	5100000	37
1901			1928	345000	11	1955	1410000	13	1982	4530000	31
1902			1929	376000	5	1956	1700000	7	1983	4230000	41
1903			1930	291000	7	1957	1830000	5	1984	4830000	36
1904			1931	166000	8	1958	1840000	12	1985	4980000	20
1905			1932	128000	11	1959	1720000	21	1986	4850000	28
1906			1933	229000	8	1960	2020000	17	1987	4600000	35
1907			1934	286000	6	1961	2060000	18	1988	5280000	45
1908			1935	340000	7	1962	2150000	12	1989	5560000	34
1909			1936	455000	8	1963	2150000	18	1990	5120000	29
1910			1937	519000	16	1964	2460000	9	1991	4300000	49
1911			1938	456000	16	1965	2770000	26	1992	4120000	44
1912			1939	577000	15	1966	2840000	19	1993	4180000	32
1913	171000	4	1940	616000	8	1967	3170000	23	1994	3750000	24
1914	121000	3	1941	698000	6	1968	3640000	19	1995	4040000	51
1915	163000	0	1942	883000	9	1969	3890000	13	1996	4180000	46
1916	201000	0	1943	1040000	9	1970	4190000	16	1997	4180000	53
1917	279000	1	1944	1040000	0	1971	4760000	21	1998	4430000	77
1918	313000	1	1945	674000	10	1972	4530000	24	1999	4300000	67
1919	196000	2	1946	524000	8	1973	4580000	28	2000	4450000	74
1920	264000	3	1947	655000	3	1974	4860000	19	2001	4590000	79
1921	92500	8	1948	795000	6	1975	4520000	22	2002	4450000	79
1922	208000	1	1949	710000	10	1976	4320000	21	2003	4850000	122
1923	215000	12	1950	844000	19	1977	4380000	38	2004	5230000	89
1924	255000	5	1951	1030000	7	1978	4670000	35	2005	5280000	101
1925	263000	9	1952	1180000	10	1979	4610000	34	2006	5330000	95
1926	310000	8	1953	1210000	5	1980	5010000	31	2007	5690000	84

Table A3.30. Correlation Eq.(A1.1) terms calculated from Table A3.29 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
233165500	2321	9.38E+14	117819	9.19E+09	3.654E+14	61113.31	3.5E+09	0.739928	54.74942

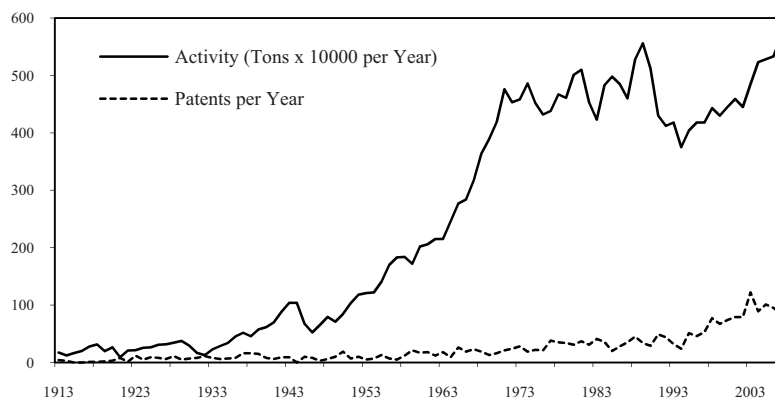


Figure A3.59. Fluorspar Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

³⁵ Fluorite or a mineral form of calcium fluoride [128].

³⁶ Activity represents world production of fluorspar, defined at usgs.gov as "...data for the years 1913 to the most recent represent the total estimated quantities of fluorspar that were produced annually throughout the world. Data were recorded from the MR [*Mineral resources of the United States*] and the MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

³⁷ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Fluorspar or fluorite were used as keywords found in the patent title or abstract by year of publication.

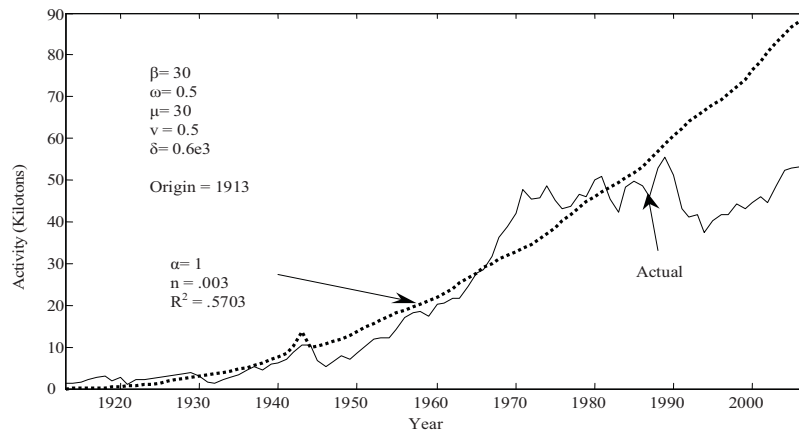


Figure A3.60. USGS World Fluorspar Production. World fluorspar production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters.

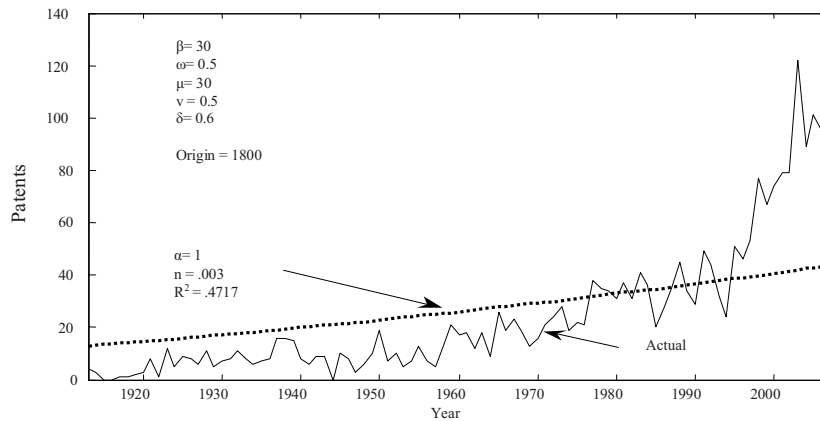


Figure A3.61. EPO Worldwide Patent Search: Fluorspar or Fluorite in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

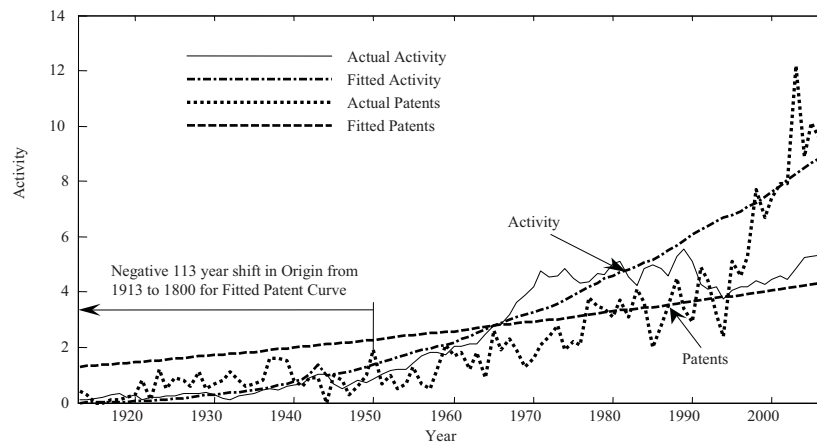


Figure A3.62. Fluorspar Best-Fit Activity and Patents. Illustrates best-fit origin shift.

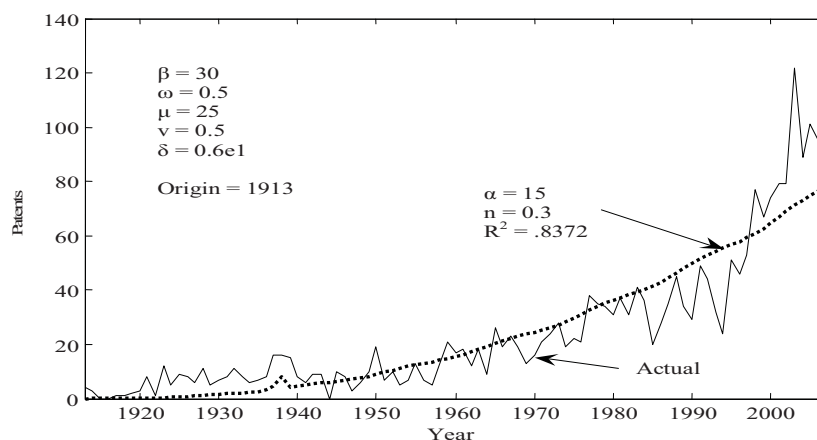


Figure A3.63. Fluorspar Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.31 Gold Activity³⁸ and Patents³⁹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	386	54	1927	597	70	1954	965	65	1981	1280	634
1901	395	45	1928	603	73	1955	947	90	1982	1340	752
1902	451	50	1929	609	87	1956	978	100	1983	1400	867
1903	496	53	1930	648	81	1957	1020	11	1984	1460	866
1904	526	48	1931	695	84	1958	1050	108	1985	1530	898
1905	575	45	1932	754	76	1959	1130	97	1986	1610	1072
1906	608	42	1933	793	80	1960	1190	180	1987	1660	1071
1907	623	45	1934	841	82	1961	1230	167	1988	1870	1204
1908	668	44	1935	924	91	1962	1290	188	1989	2010	1348
1909	687	39	1936	1030	75	1963	1340	242	1990	2180	1442
1910	689	51	1937	1100	77	1964	1390	262	1991	2160	1418
1911	699	34	1938	1170	89	1965	1440	324	1992	2260	1530
1912	705	42	1939	1230	81	1966	1450	321	1993	2280	1315
1913	694	46	1940	1310	52	1967	1420	364	1994	2260	1410
1914	663	41	1941	1080	57	1968	1440	301	1995	2230	1280
1915	704	22	1942	1120	31	1969	1450	360	1996	2290	1250
1916	685	30	1943	896	25	1970	1480	404	1997	2450	1268
1917	631	22	1944	813	31	1971	1450	412	1998	2500	1466
1918	578	24	1945	762	34	1972	1390	481	1999	2570	1533
1919	550	28	1946	860	35	1973	1350	454	2000	2590	1763
1920	507	32	1947	900	58	1974	1250	381	2001	2600	1746
1921	498	38	1948	932	52	1975	1200	454	2002	2550	1891
1922	481	42	1949	964	66	1976	1210	436	2003	2540	1958
1923	554	52	1950	879	66	1977	1210	427	2004	2420	1869
1924	592	52	1951	883	64	1978	1210	521	2005	2470	1710
1925	591	59	1952	868	84	1979	1210	522	2006	2430	1643
1926	602	56	1953	864	75	1980	1220	639	2007	2380	1839

³⁸ Activity represents world production of gold, defined at usgs.gov as "...World gold production data for the years 1900–26 are from reported estimates by Ridgeway (1929). World gold production data for the years 1927 to the most recent are from the MYB in the "Salient gold statistics" and "Gold: World production by country" tables. Updated values for world gold production for the years 1929–50 reflect revised estimates by the USGS gold commodity specialist for some countries." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

³⁹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Gold was used as the keyword found in the patent title or abstract by year of publication.

Table A3.32. Correlation Eq.(A1.1) terms calculated from Table A3.31 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
131193	48266	2.02E+08	57810854	9.5481495E+07	42538212	36240421	36850372	0.938546	88.08688898

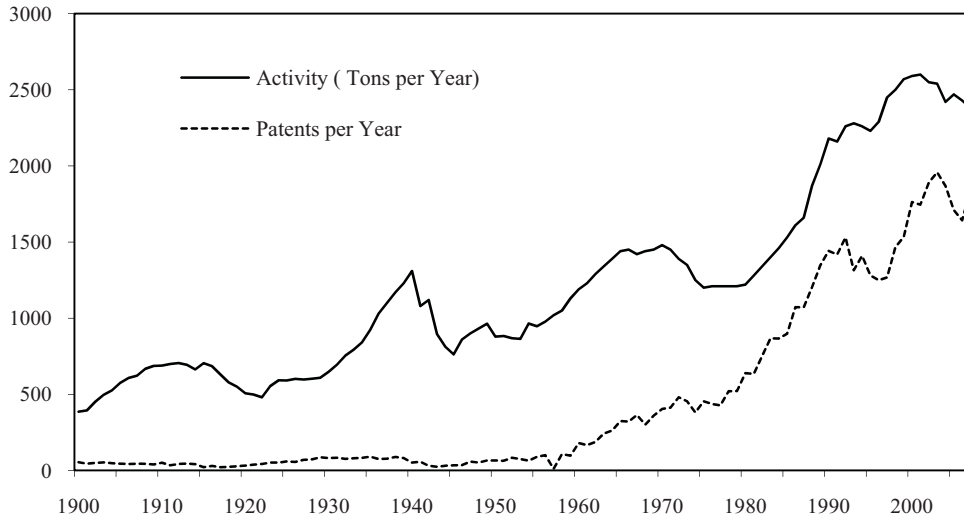


Figure A3.64. Gold Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

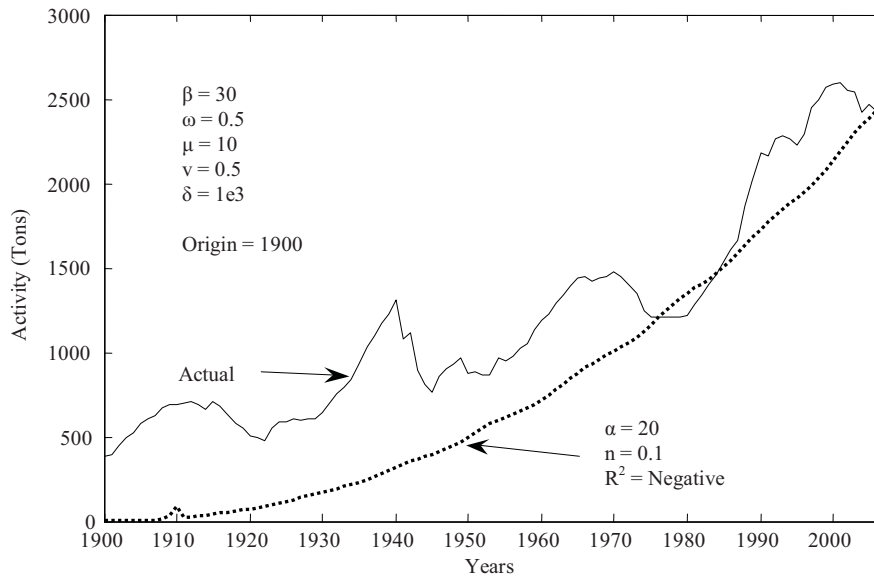


Figure A3.65. USGS World Gold Production. World gold production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. R² is negative possibly indicating Stage IV. No best-fit for the patent data was obtainable.

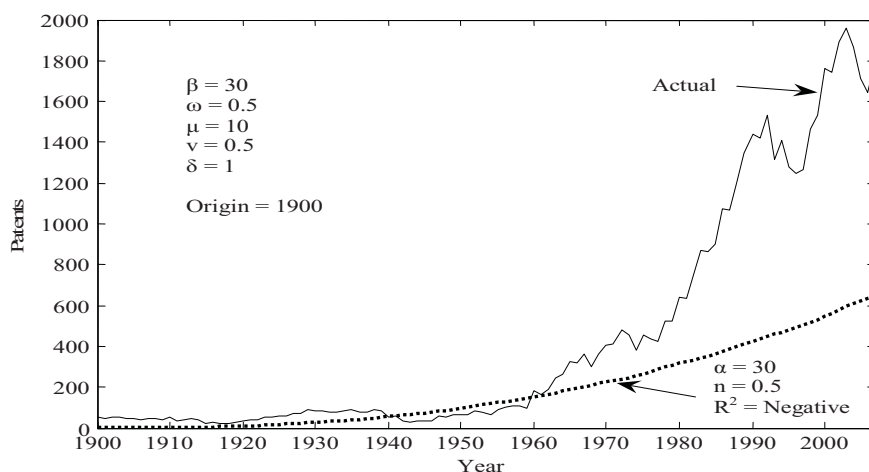


Figure A3.66. Gold Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.33. Graphite⁴⁰ Activity⁴¹ and Patents⁴²

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	81600	29	1927	154000	100	1954	168000	177	1981	589000	1070
1901	77100	20	1928	150000	93	1955	272000	217	1982	562000	1185
1902	81600	24	1929	150000	93	1956	263000	217	1983	604000	1303
1903	77100	29	1930	118000	100	1957	372000	271	1984	625000	1305
1904	77100	38	1931	77100	125	1958	318000	253	1985	584000	1307
1905	90700	52	1932	72600	119	1959	372000	312	1986	625000	1407
1906	104000	46	1933	86200	102	1960	435000	447	1987	643000	1440
1907	109000	44	1934	109000	90	1961	413000	403	1988	575000	1418
1908	95300	51	1935	145000	124	1962	535000	433	1989	1010000	1744
1909	95300	43	1936	150000	103	1963	679000	402	1990	946000	1566
1910	95300	33	1937	159000	123	1964	620000	439	1991	771000	1528
1911	109000	43	1938	177000	122	1965	607000	456	1992	670000	1643
1912	118000	42	1939	222000	116	1966	484000	423	1993	648000	1447
1913	136000	46	1940	254000	92	1967	358000	536	1994	517000	1509
1914	104000	28	1941	231000	66	1968	437000	520	1995	584000	1517
1915	113000	36	1942	272000	57	1969	376000	498	1996	555000	1546
1916	172000	20	1943	277000	67	1970	393000	608	1997	685000	1450
1917	209000	29	1944	263000	55	1971	394000	595	1998	651000	1666
1918	181000	25	1945	136000	56	1972	361000	671	1999	692000	1839
1919	122000	36	1946	95300	78	1973	395000	584	2000	846000	1940
1920	118000	54	1947	122000	106	1974	497000	549	2001	816000	1796
1921	90700	81	1948	163000	137	1975	451000	568	2002	932000	2004
1922	104000	78	1949	168000	117	1976	449000	621	2003	999000	1966
1923	95300	70	1950	159000	103	1977	493000	640	2004	1020000	2072
1924	95300	79	1951	195000	122	1978	534000	824	2005	1040000	2078
1925	122000	83	1952	177000	166	1979	626000	609	2006	1020000	1835
1926	136000	67	1953	168000	132	1980	597000	1008	2007	1110000	1922

⁴⁰ Natural Graphite [78].

⁴¹ Activity represents world production of graphite, defined at usgs.gov as follows. "...were reported in the MR [*Minerals Resources of the United States*] and the MYB [*Minerals Yearbook*]." (usgs.gov) Data is in metric tons, as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁴² Patents are total patents in a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Graphite was used as the keyword found in the patent title or abstract by year of publication.

Table A3.34. Correlation Eq.(A1.1) terms calculated from Table A3.33 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
39677600	60774	2.3E+13	79940074	4.06E+10	8.438E+12	45741194	1.82E+10	0.928652	86.23946

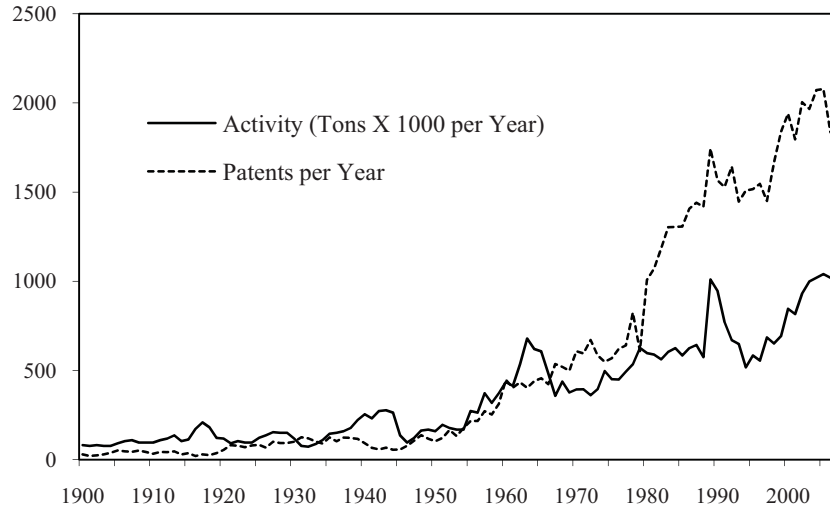


Figure A3.67. Graphite Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

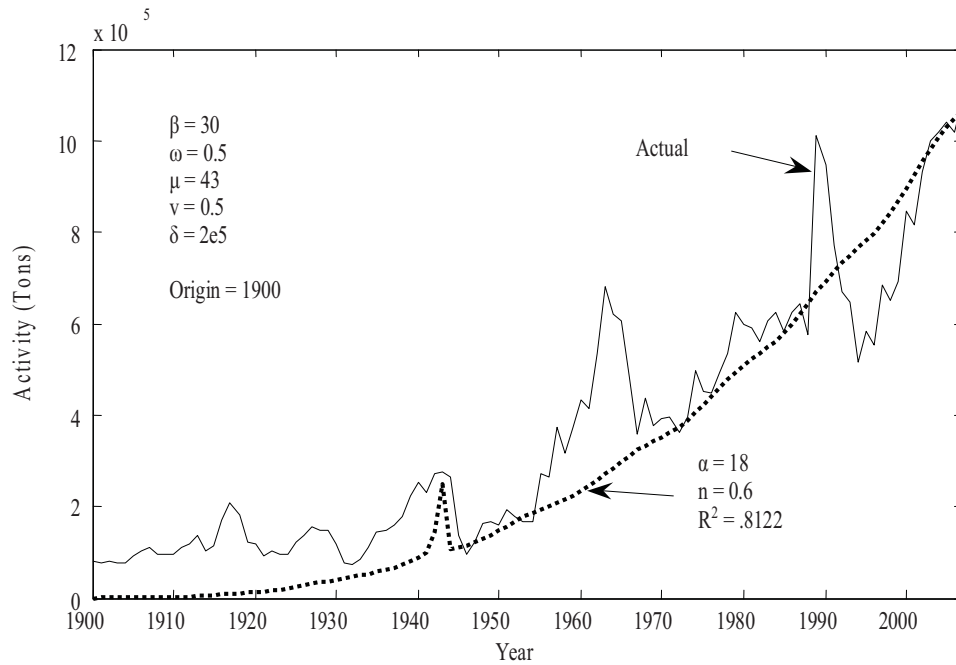


Figure A3.68. USGS World Graphite Production. World graphite production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

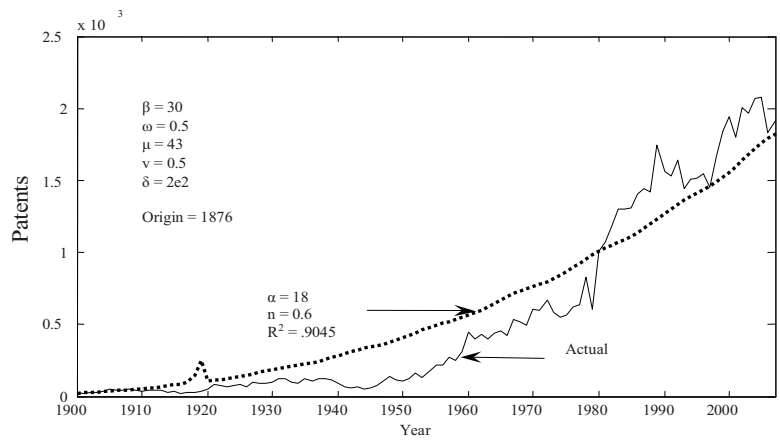


Figure A3.69. EPO worldwide Patent Search: Graphite in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

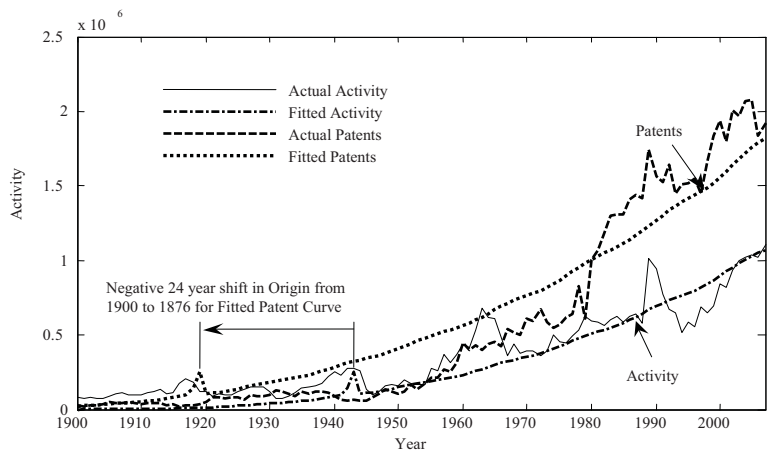


Figure A3.70. Graphite Best-Fit Activity and Patents. Illustrates best-fit origin shift.

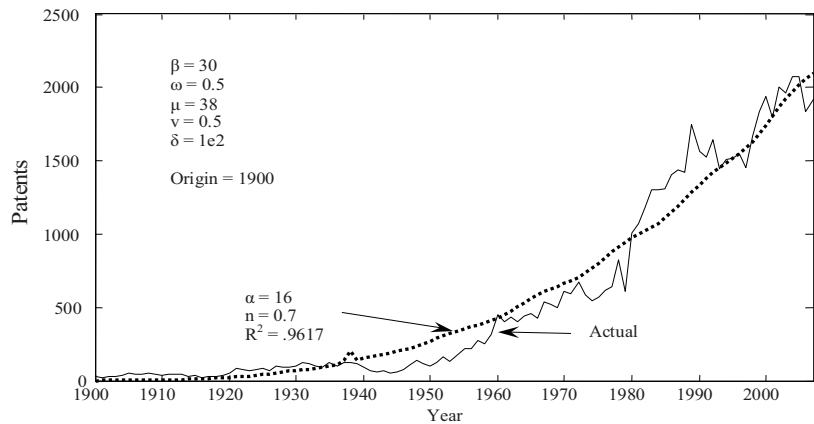


Figure A3.71. Graphite Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.35 Gypsum⁴³ Activity⁴⁴ and Patents⁴⁵

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y
1900			1927	11200000	35	1954	28000000	31	1981	76200000	374
1901			1928	11800000	19	1955	32100000	26	1982	72500000	342
1902			1929	12500000	37	1956	33500000	31	1983	80700000	385
1903			1930	11900000	46	1957	34200000	40	1984	85800000	337
1904			1931	9400000	46	1958	37800000	35	1985	87000000	336
1905			1932	7800000	49	1959	43100000	46	1986	88200000	317
1906			1933	7400000	54	1960	40000000	37	1987	93000000	325
1907			1934	7900000	35	1961	40500000	33	1988	101000000	355
1908			1935	8300000	45	1962	43500000	40	1989	104000000	427
1909			1936	9400000	31	1963	45500000	42	1990	104000000	409
1910			1937	7790000	30	1964	46800000	50	1991	100000000	421
1911			1938	6030000	44	1965	48000000	64	1992	98800000	531
1912			1939	8030000	23	1966	48700000	48	1993	97200000	451
1913			1940	7940000	29	1967	46200000	75	1994	96300000	465
1914			1941	8960000	14	1968	49400000	78	1995	98400000	488
1915			1942	9350000	19	1969	52200000	66	1996	104000000	522
1916			1943	8480000	13	1970	51600000	64	1997	107000000	629
1917			1944	8400000	17	1971	53100000	119	1998	104000000	678
1918			1945	9800000	12	1972	57600000	126	1999	109000000	703
1919			1946	14400000	17	1973	61500000	94	2000	108000000	735
1920			1947	16500000	24	1974	61400000	132	2001	105000000	778
1921			1948	21200000	17	1975	59200000	142	2002	111000000	764
1922			1949	19000000	13	1976	66100000	219	2003	114000000	751
1923			1950	22600000	19	1977	74500000	328	2004	120000000	774
1924	9700000	13	1951	18400000	25	1978	77800000	420	2005	122000000	781
1925	10700000	26	1952	24100000	18	1979	80400000	336	2006	125000000	772
1926	11300000	28	1953	25400000	21	1980	78400000	384	2007	152400000	784

Table A3.36. Correlation Eq.(A1.1) terms calculated from Table A3.35 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
4.521E+09	18559	3.73E+17	9503115	1.77E+12	1.292E+17	5402681	7.7E+11	0.921142	84.85017

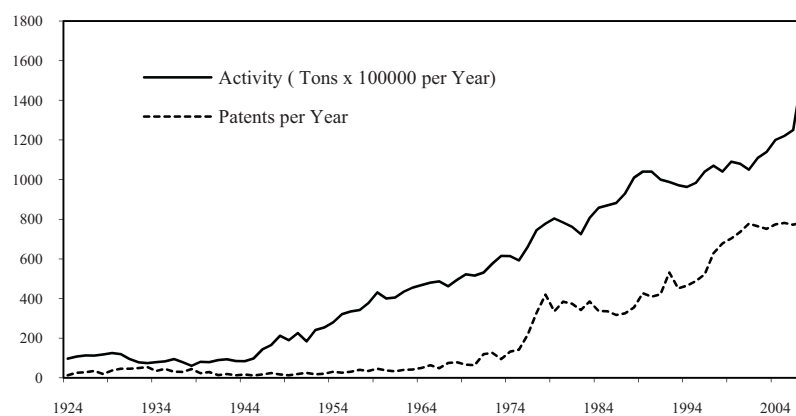


Figure A3.72. Gypsum Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁴³ Monoclinic mineral form of hydrated calcium sulphate.

⁴⁴ Activity represents world production of gypsum, defined at usgs.gov as "...mine production of crude gypsum. Data are not available prior to 1924. Data are recorded in the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁴⁵ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Gypsum was used as the keyword found in the patent title or abstract by year of publication.

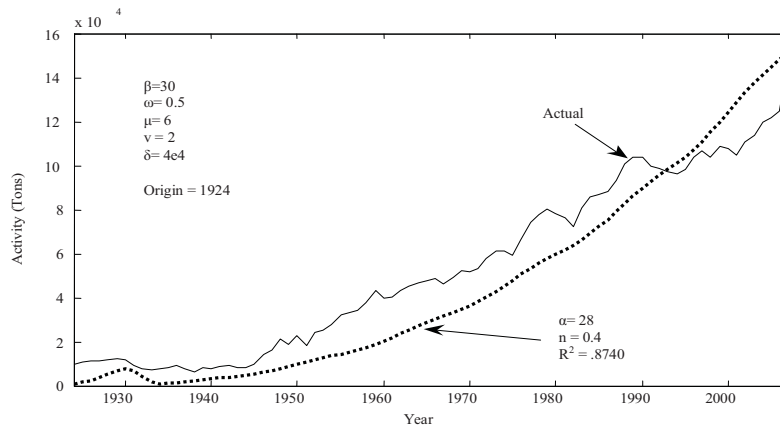


Figure A3.73. USGS World Gypsum Production. World gypsum production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

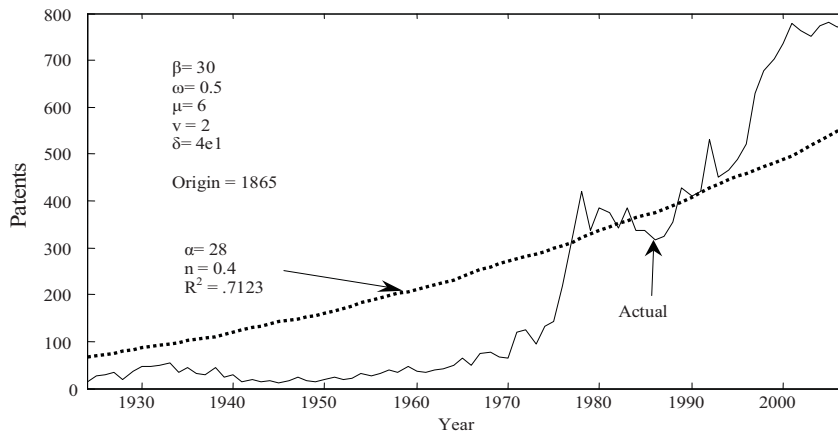


Figure A3.74. EPO Worldwide Patent Search: Gypsum in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

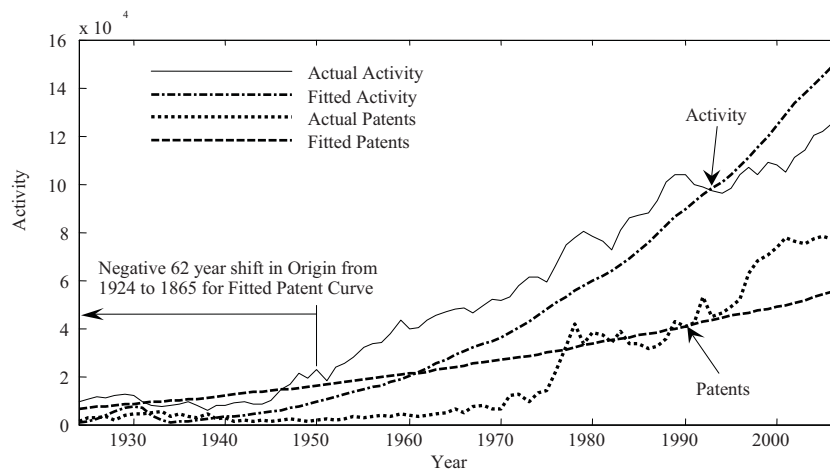


Figure A3.75. Graphite Best-Fit Activity and Patents. Illustrates best-fit origin shift.

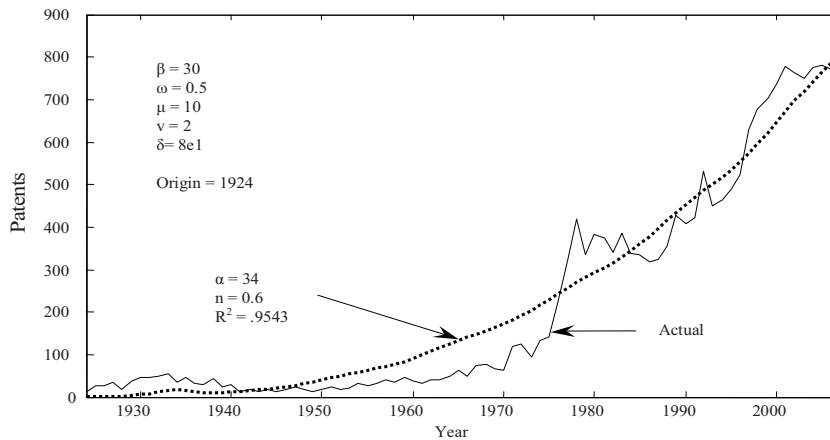


Figure A3.76. Gypsum Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.37. Helium Activity⁴⁶ and Patents⁴⁷

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	914	27	1981	5580	198
1901			1928			1955	1060	50	1982	1830	270
1902			1929			1956	1170	58	1983	3480	256
1903			1930			1957	1400	49	1984	8570	309
1904			1931			1958	1600	64	1985	9750	371
1905			1932			1959	2290	58	1986	10000	355
1906			1933			1960	3080	100	1987	11600	429
1907			1934			1961	3490	90	1988	13300	406
1908			1935	49	21	1962	3420	103	1989	14800	548
1909			1936	22.4	16	1963	10800	122	1990	15600	420
1910			1937	23.1	15	1964	19400	110	1991	15900	426
1911			1938	29.2	21	1965	21000	147	1992	16900	561
1912			1939	30.1	17	1966	22100	130	1993	16900	453
1913			1940	45.3	10	1967	22700	169	1994	17900	525
1914			1941	77.5	3	1968	22500	157	1995	18800	396
1915			1942	159	6	1969	22500	174	1996	18800	397
1916			1943	558	2	1970	22200	173	1997	23400	376
1917			1944	608	1	1971	22400	165	1998	22700	484
1918			1945	454	10	1972	20200	203	1999	22900	478
1919			1946	279	10	1973	16000	212	2000	19800	571
1920			1947	337	12	1974	4900	179	2001	17900	564
1921			1948	303	14	1975	5870	177	2002	18500	665
1922			1949	264	10	1976	7120	156	2003	24400	594
1923			1950	390	13	1977	7830	140	2004	26100	555
1924			1951	537	19	1978	8400	176	2005	27100	481
1925			1952	693	31	1979	8890	158	2006	28100	436
1926			1953	772	31	1980	7560	181	2007	28900	428

⁴⁶ Activity represents world production of helium, defined at usgs.gov as follows. "...World production data for the years 1935–71 were recorded from the MYB. World production data for the years 1972 to the most recent were recorded from the MCS. World production data for the years 1935 to the most recent represent the summed quantity of total U.S. helium production and the total estimated production capacity of all other helium-producing countries. For the years 1935–62, world production is equal to U.S. production." Data is in metric tons, as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁴⁷ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Helium was used as the keyword found in the patent title or abstract by year of publication.

Table A3.38. Correlation Eq.(A1.1) terms calculated from Table A3.37 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
755934.6	15742	1.43E+10	6148154	2.62E+08	6.425E+09	2753489	99225238	0.745984	55.64915

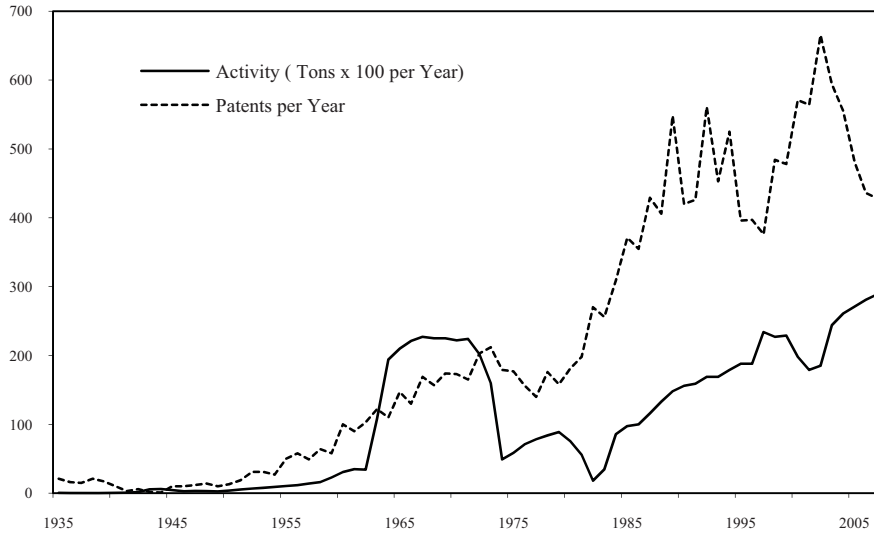


Figure A3.77. Helium Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

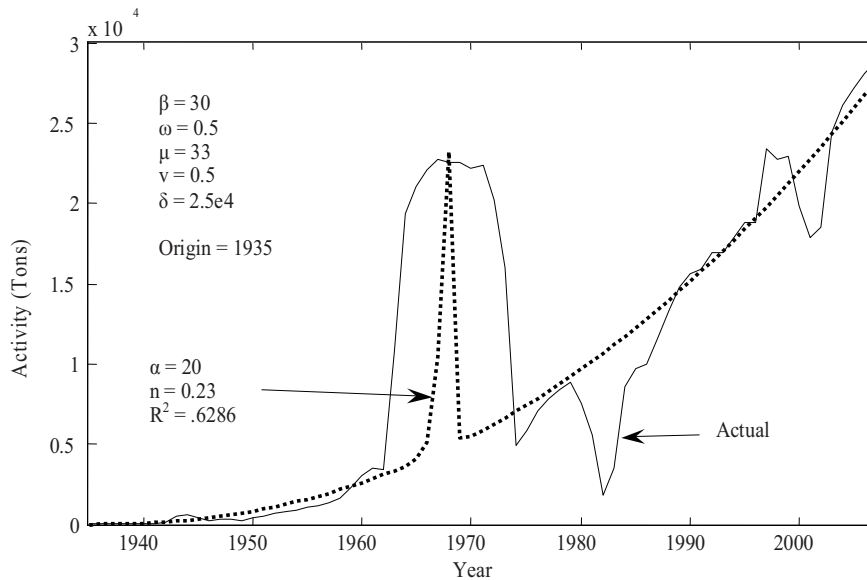


Figure A3.78. USGS World Helium Production. World helium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

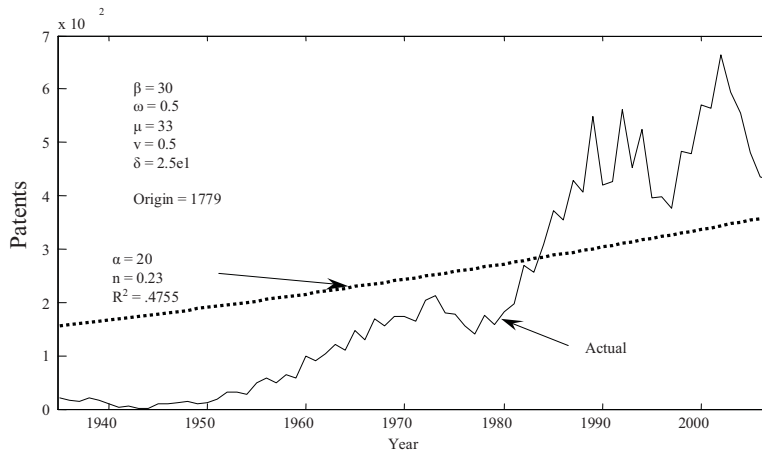


Figure A3.79. EPO Worldwide Patent Search: Helium in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

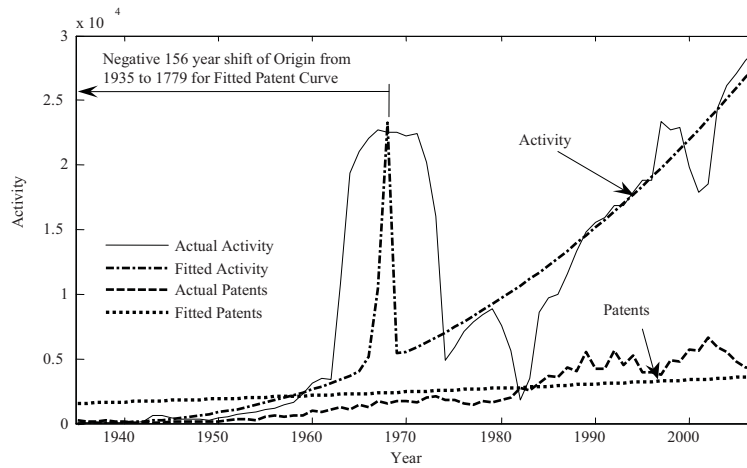


Figure A3.80. Helium Best-Fit Activity and Patents. Illustrates best-fit origin shift.

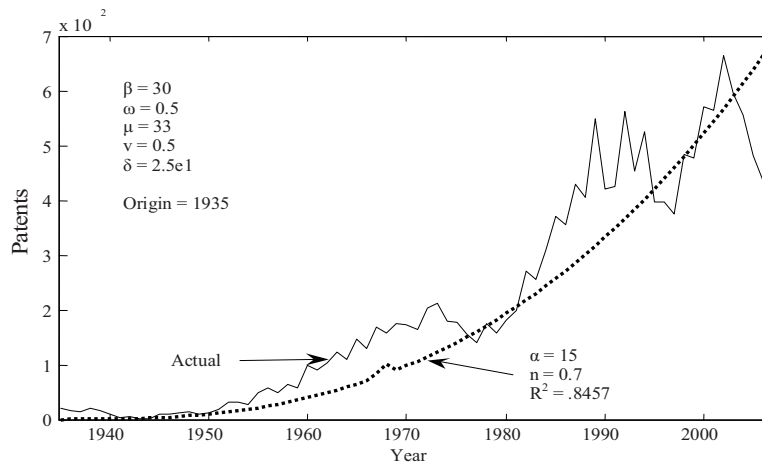


Figure A3.81. Helium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.39 Hydraulic Cement⁴⁸ Activity⁴⁹ and Patents⁵⁰

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (pat)
1900			1927	67800000	13	1954	194900000	17	1981	886700000	73
1901			1928	72200000	11	1955	217300000	17	1982	887400000	102
1902			1929	74900000	16	1956	235400000	10	1983	916600000	96
1903			1930	72300000	12	1957	246900000	13	1984	941100000	124
1904			1931	62100000	15	1958	262500000	13	1985	959400000	103
1905			1932	49300000	21	1959	294300000	21	1986	1.008E+09	92
1906			1933	48200000	22	1960	316500000	26	1987	1.053E+09	114
1907			1934	58300000	20	1961	333200000	18	1988	1.118E+09	99
1908			1935	65400000	24	1962	358500000	19	1989	1.042E+09	152
1909			1936	62800000	12	1963	378000000	22	1990	1.043E+09	150
1910			1937	82700000	18	1964	415600000	20	1991	1.185E+09	183
1911			1938	85900000	23	1965	433400000	32	1992	1.123E+09	174
1912			1939	93000000	16	1966	464200000	22	1993	1.291E+09	207
1913			1940	81000000	7	1967	479800000	30	1994	1.37E+09	213
1914			1941	88000000	9	1968	515200000	28	1995	1.445E+09	212
1915			1942	80900000	7	1969	543100000	26	1996	1.493E+09	154
1916			1943	71200000	9	1970	571800000	29	1997	1.547E+09	197
1917			1944	54900000	18	1971	590000000	37	1998	1.54E+09	185
1918			1945	49500000	6	1972	661000000	34	1999	1.6E+09	226
1919			1946	72500000	4	1973	702000000	39	2000	1.66E+09	203
1920			1947	85800000	4	1974	703200000	49	2001	1.75E+09	200
1921			1948	102000000	13	1975	702200000	53	2002	1.85E+09	219
1922			1949	115000000	12	1976	735400000	65	2003	2.02E+09	211
1923			1950	133000000	11	1977	797100000	68	2004	2.19E+09	228
1924			1951	149000000	5	1978	853000000	77	2005	2.35E+09	207
1925			1952	161000000	6	1979	872400000	39	2006	2.55E+09	209
1926	62400000	11	1953	178000000	12	1980	883100000	66	2007	2.77E+09	197

Table A3.40. Correlation Eq.(A1.1) terms calculated from Table A3.39 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
5.573E+10	5777	7.39E+19	867369	7.71E+12	3.604E+19	460372.3	3.79E+12	0.929922	86.4755

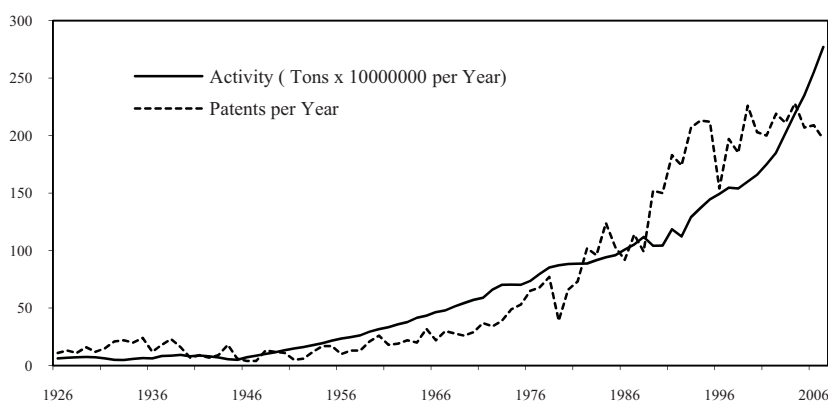


Figure A3.82. Hydraulic Cement Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁴⁸ Portland, natural, masonry, and slag or pozzolanic cement [78].

⁴⁹ Activity represents world production of hydraulic cement, defined at usgs.gov as "...recorded from the MYB [*Minerals Yearbook*] and MR [*Minerals Resources of the United States*]. World production statistics were not available from 1900-1925" Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁵⁰ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Hydraulic and cement were used as the keywords found in the patent title or abstract by year of publication.

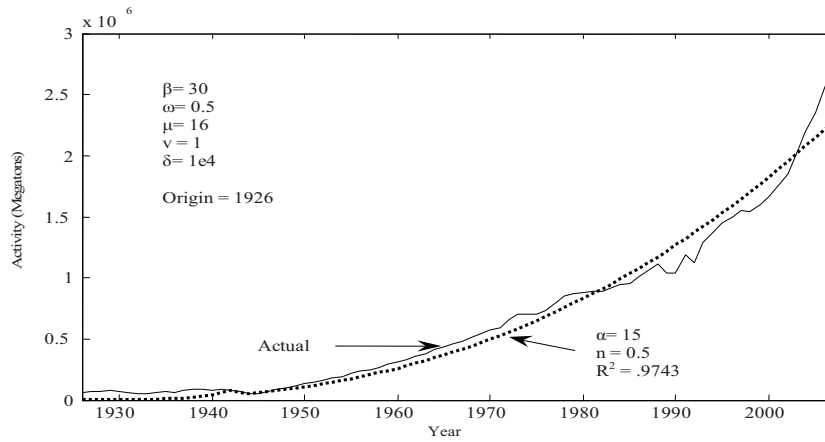


Figure A3.83. USGS World Hydraulic Cement Production. World hydraulic cement production scaled in metric megatons with actual and best-fit curves and common pattern equation parameters.

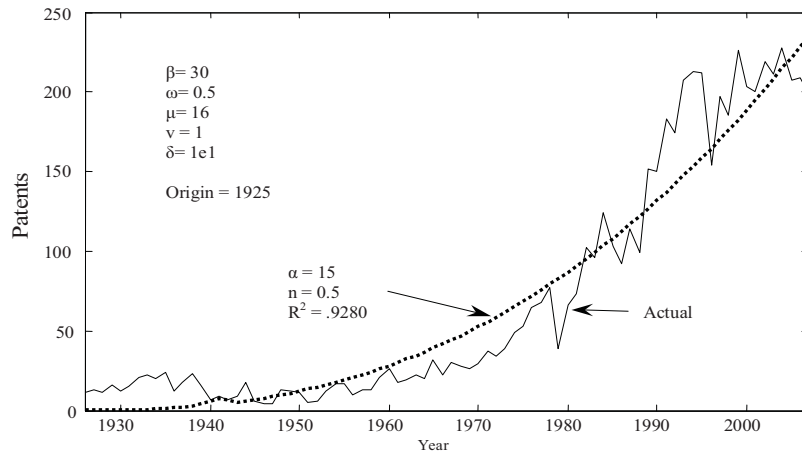


Figure A3.84. EPO Worldwide Patent Search: Hydraulic Cement in Title or Abstract by date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

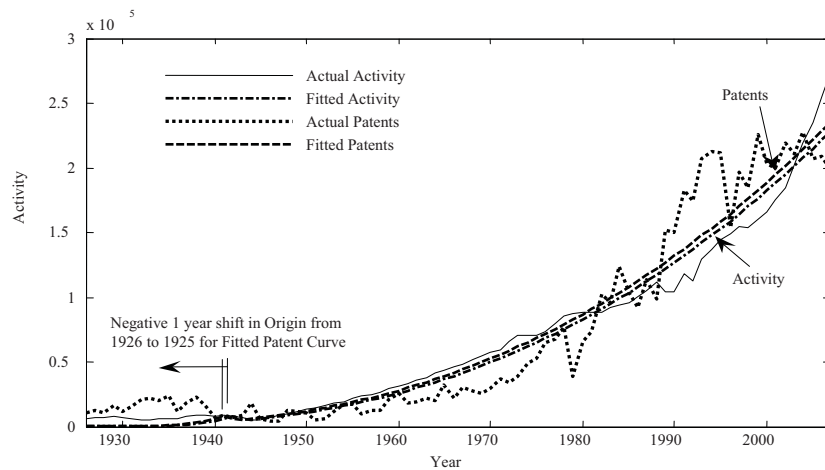


Figure A3.85. Hydraulic Cement Best-Fit activity and Patents. Illustrates best-fit origin shift.

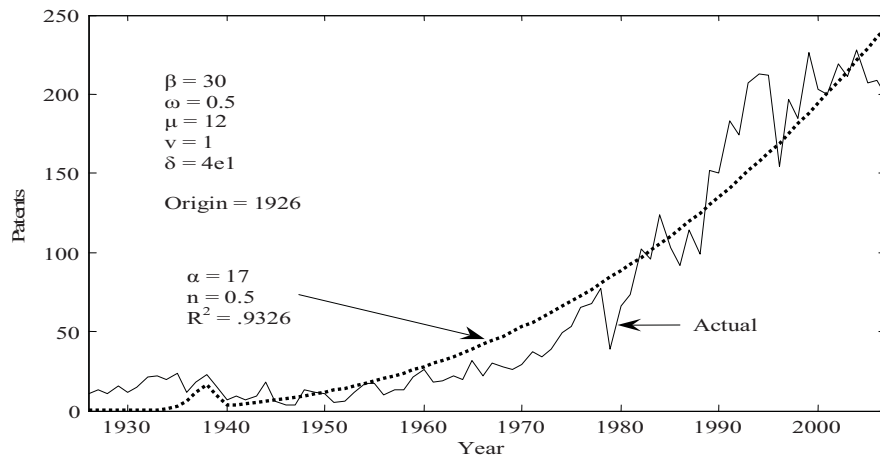


Figure A3.86. Hydraulic Cement Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.41 Iodine Activity⁵¹ and Patents⁵²

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981	12000	443
1901			1928			1955			1982	12300	452
1902			1929			1956			1983	12500	466
1903			1930			1957			1984	12400	387
1904			1931			1958			1985	12800	482
1905			1932			1959			1986	13000	496
1906			1933			1960	3030	143	1987	12700	490
1907			1934			1961	3360	147	1988	14900	587
1908			1935			1962	3410	131	1989	16300	642
1909			1936			1963	3580	171	1990	16000	643
1910			1937			1964	4190	207	1991	17300	538
1911			1938			1965	4480	254	1992	16500	626
1912			1939			1966	5560	177	1993	16100	602
1913			1940			1967	5250	232	1994	14300	654
1914			1941			1968	5290	195	1995	13400	669
1915			1942			1969	7070	188	1996	14100	681
1916			1943			1970	8260	211	1997	15700	689
1917			1944			1971	9360	167	1998	18600	795
1918			1945			1972	9740	228	1999	18400	701
1919			1946			1973	10900	209	2000	19500	823
1920			1947			1974	10400	184	2001	20700	679
1921			1948			1975	10800	205	2002	21000	779
1922			1949			1976	11000	265	2003	24600	858
1923			1950			1977	10300	296	2004	24800	732
1924			1951			1978	10400	282	2005	26500	773
1925			1952			1979	11100	242	2006	26700	719
1926			1953			1980	11600	482	2007	25700	738

⁵¹ Activity represents world production of iodine, defined at usgs.gov as “World production data for the years 1960-75, are reported in the mine production table of the CDS [*Commodity Data Summaries*]. Data for the years 1976 to the most recent are reported in the world production table of the MYB [*Minerals Yearbook*]. Excludes production in the U.S. in 2006.” Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁵² Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Iodine was used as a keyword found in the patent title or abstract by year of publication.

Table A3.42. Correlation Eq.(A1.1) terms calculated from Table A3.41 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
614500	21168	1.01E+10	12351438	3.46E+08	2.228E+09	3016350	74640680	0.910526	82.90567

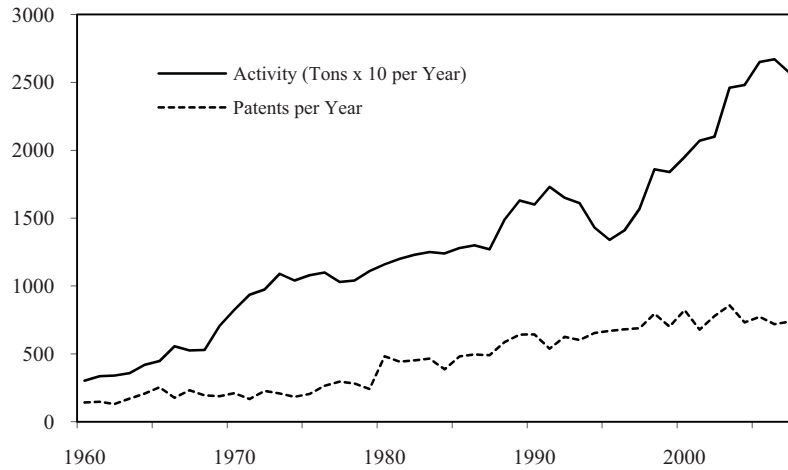


Figure A3.87. Iodine Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

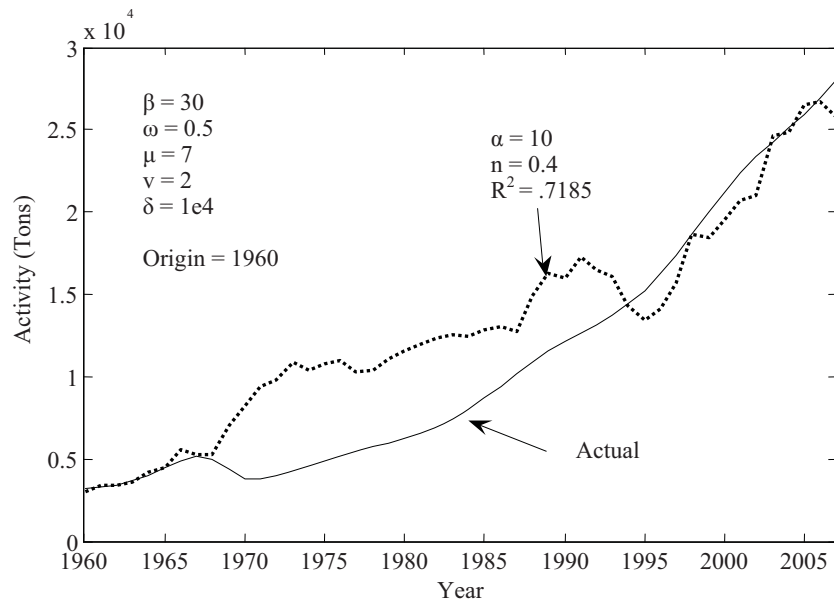


Figure A3.88. USGS World Iodine Production. World iodine production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

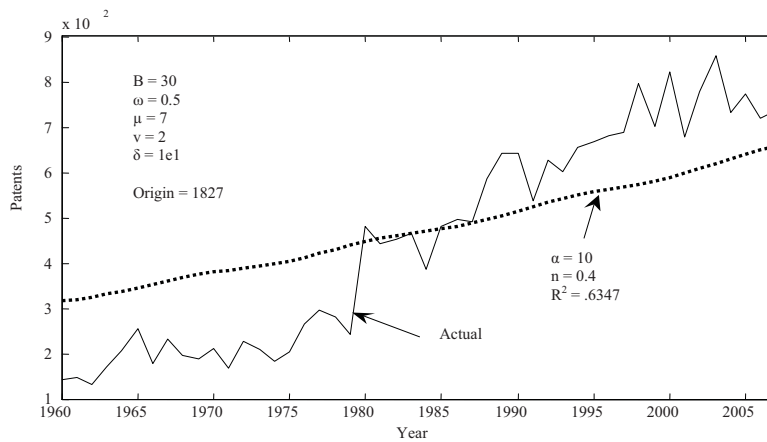


Figure A3.89. EPO Worldwide Patent Search: Iodine in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

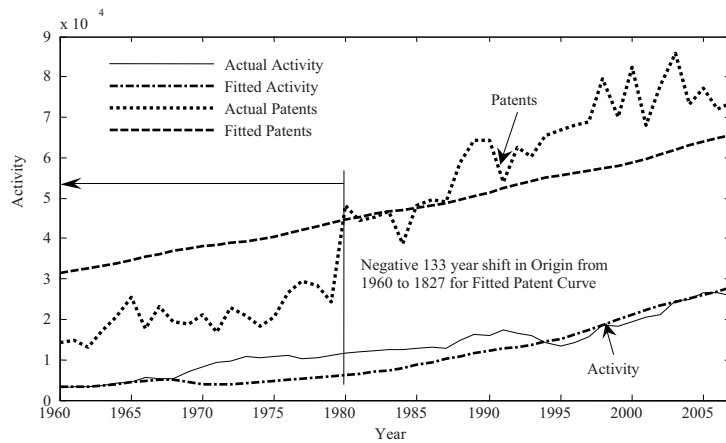


Figure A3.90. Iodine Best-Fit Activity and Patents. Illustrates best-fit origin shift.

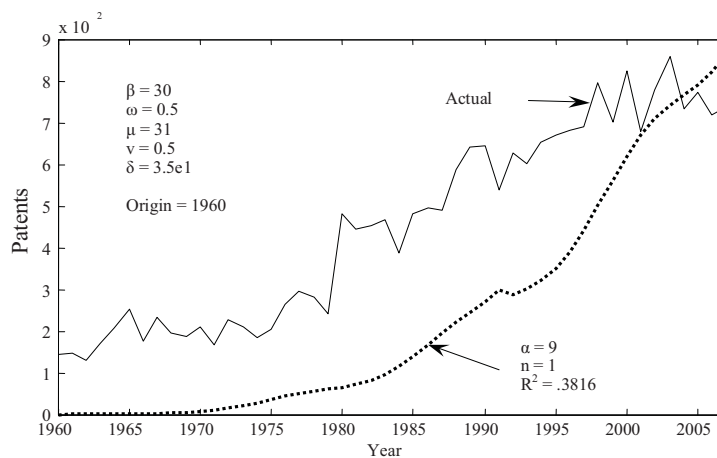


Figure A3.91. Iodine Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.43 Iron Activity⁵³ and Patents⁵⁴

Year	x (activity)	y pat.	Year	x (activity)	y (patent)	Year	x (activity)	y (pat.)	Yr.	x (activity)	y (pat.)
1900			1927	171000000	922	1954	405000000	954	1981	858000000	6201
1901			1928	174000000	909	1955	369000000	1088	1982	781000000	7271
1902			1929	201000000	964	1956	395000000	1320	1983	740000000	7605
1903			1930	179000000	1130	1957	434000000	1215	1984	829000000	7745
1904	95500000	717	1931	119000000	1281	1958	405000000	1123	1985	861000000	8070
1905	116000000	662	1932	76200000	1121	1959	439000000	1103	1986	864000000	8961
1906	100000000	590	1933	91200000	964	1960	522000000	1525	1987	903000000	8758
1907	135000000	598	1934	120000000	910	1961	503000000	1341	1988	967000000	9147
1908	109000000	580	1935	138000000	965	1962	508000000	1337	1989	1.01E+09	10130
1909	126000000	548	1936	170000000	876	1963	523000000	1466	1990	983000000	9728
1910	142000000	521	1937	212000000	912	1964	583000000	1469	1991	956000000	9348
1911	133000000	474	1938	162000000	999	1965	621000000	1741	1992	925000000	10660
1912	151000000	464	1939	204000000	828	1966	636000000	1482	1993	953000000	9741
1913	177000000	555	1940	204000000	698	1967	623000000	1837	1994	992000000	9994
1914	118000000	442	1941	220000000	493	1968	679000000	1772	1995	1.03E+09	9689
1915	116000000	351	1942	235000000	434	1969	713000000	1638	1996	1.02E+09	9529
1916	139000000	256	1943	231000000	368	1970	769000000	2067	1997	1.07E+09	9576
1917	142000000	265	1944	203000000	376	1971	787000000	2198	1998	1.05E+09	10732
1918	127000000	288	1945	162000000	445	1972	778000000	2446	1999	1.02E+09	10700
1919	110000000	433	1946	154000000	455	1973	846000000	2341	2000	1.07E+09	11910
1920	124000000	550	1947	187000000	535	1974	898000000	2283	2001	1.04E+09	11425
1921	73000000	746	1948	219000000	821	1975	902000000	2528	2002	1.1E+09	12041
1922	104000000	723	1949	223000000	904	1976	899000000	2790	2003	1.21E+09	11430
1923	136000000	710	1950	251000000	680	1977	841000000	3792	2004	1.36E+09	11736
1924	130000000	699	1951	294000000	769	1978	847000000	4042	2005	1.54E+09	11190
1925	151000000	823	1952	297000000	1106	1979	903000000	4149	2006	1.82E+09	11033
1926	155000000	800	1953	338000000	732	1980	891000000	6089	2007	2.03E+09	11840

Table A3.44. Correlation Eq.(A1.1) terms calculated from Table A3.43 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
5.485E+10	357718	4.74E+19	2.87E+09	3.41E+14	1.852E+19	1.64E+09	1.52E+14	0.874126	76.40968

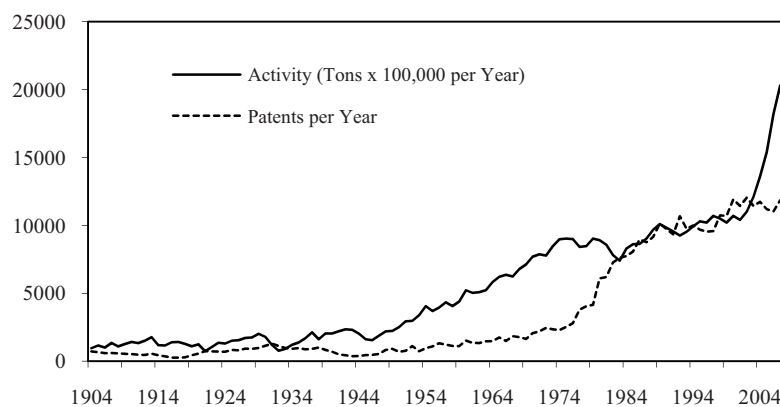


Figure A3.92. Iron Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁵³ Activity represents world production of iron, defined at usgs.gov as "...the world production of iron ore, iron ore concentrates, and iron ore agglomerates. For the years 1913–22, world production is reported as "production in principal countries." A graph of the time series for world production gives a smooth curve when the category name changes, indicating that major producers were included. World production data were recorded from the MR [*Minerals Resources of the United States*] and the MYB [*Minerals Yearbook*]. "Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁵⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Iron of Fe were used as keyword founds in the patent title or abstract by year of publication.

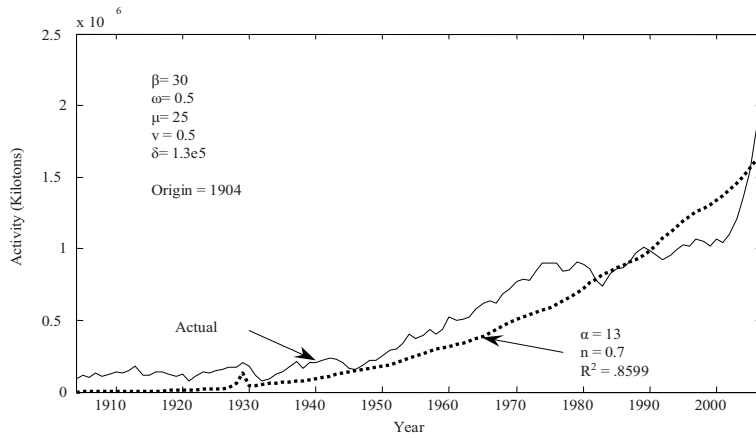


Figure A3.93. USGS World Iron Production. World iron production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters.

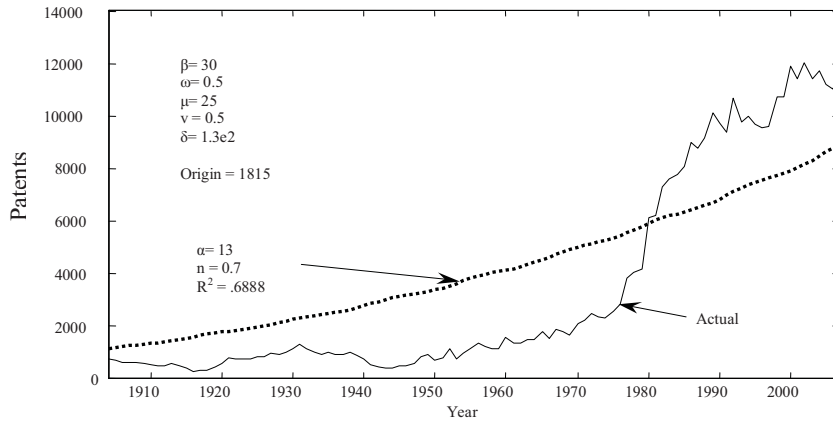


Figure A3.94. EPO Worldwide Patent Search: Iron or Fe in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

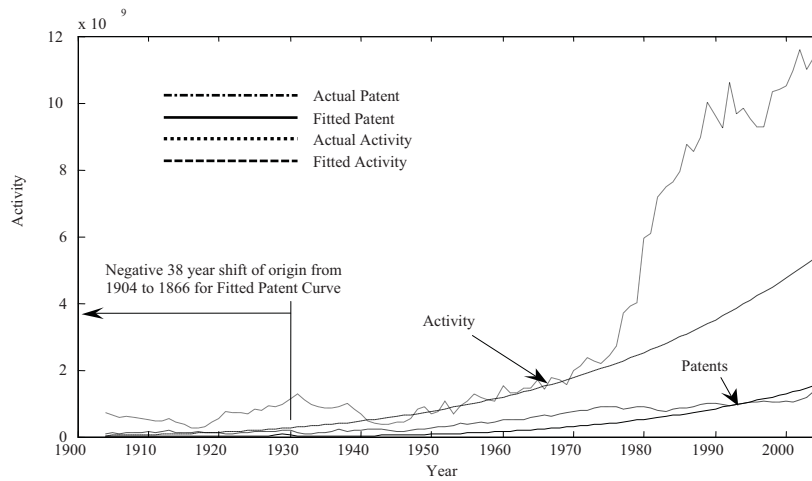


Figure A3.95. Iron Best-Fit Activity and Patents. Illustrates best-fit origin shift.

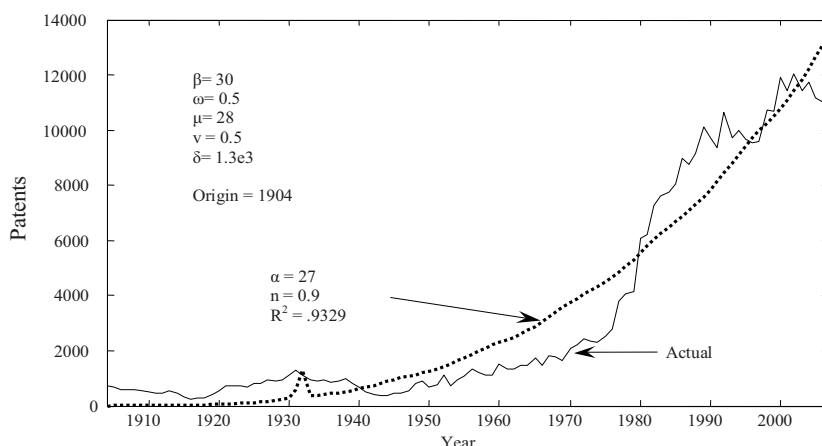


Figure A3.96. Iron Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.45 Kyanite⁵⁵ Activity⁵⁶ and Patents⁵⁷

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	73000	3	1981	330000	120
1901			1928	2300	0	1955	38000	0	1982	290000	132
1902			1929	3700	0	1956	58000	1	1983	240000	140
1903			1930	8800	1	1957	82000	1	1984	280000	122
1904			1931	3500	1	1958	76000	4	1985	340000	114
1905			1932	5700	0	1959	77000	3	1986	300000	114
1906			1933	4500	1	1960	95000	3	1987	340000	121
1907			1934	9700	0	1961	130000	3	1988	380000	133
1908			1935	20000	0	1962	130000	3	1989	410000	154
1909			1936	25000	1	1963	120000	8	1990	400000	137
1910			1937	27000	0	1964	120000	7	1991	310000	146
1911			1938	29000	1	1965	120000	8	1992	320000	142
1912			1939	15000	0	1966	130000	11	1993	270000	118
1913			1940	13000	0	1967	130000	7	1994	280000	147
1914			1941	22000	3	1968	120000	4	1995	280000	150
1915			1942	25000	1	1969	140000	6	1996	300000	163
1916			1943	23000	0	1970	200000	10	1997	350000	152
1917			1944	15000	1	1971	190000	6	1998	460000	170
1918			1945	15000	1	1972	180000	13	1999	360000	207
1919			1946	23000	4	1973	200000	12	2000	406000	209
1920			1947	43000	0	1974	180000	13	2001	424000	201
1921			1948	46000	0	1975	190000	33	2002	391000	182
1922			1949	61000	2	1976	230000	33	2003	386000	202
1923			1950	58000	2	1977	250000	29	2004	456000	211
1924			1951	73000	0	1978	240000	43	2005	450000	189
1925			1952	67000	2	1979	290000	69	2006	444000	176
1926			1953	47000	2	1980	350000	123	2007	443000	170

⁵⁵ Includes synthetic mullite and kyanite [78].

⁵⁶ Activity represents world production of kyanite, defined at usgs.gov as "...data for the years 1928-60 are from the "World Production" table in the 1960 MYB [Minerals Yearbook]. World production data for the years 1961-70 were from the "World Mine Production" table in the CDS. World production for the years 1971-2002 were from the MCS[Mineral Commodity Summaries]. Data for the years 2003 to the most recent are unpublished revisions made by the Commodity Specialist." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁵⁷ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Kyanite or aluminum silicate were used as keywords found in the patent title or abstract by year of publication.

Table A3.46. Correlation Eq.(A1.1) terms calculated from Table A3.45 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
14464200	4701	4.36E+12	712065	1.66E+09	1.748E+12	435822.5	8.07E+08	0.92419	85.4127

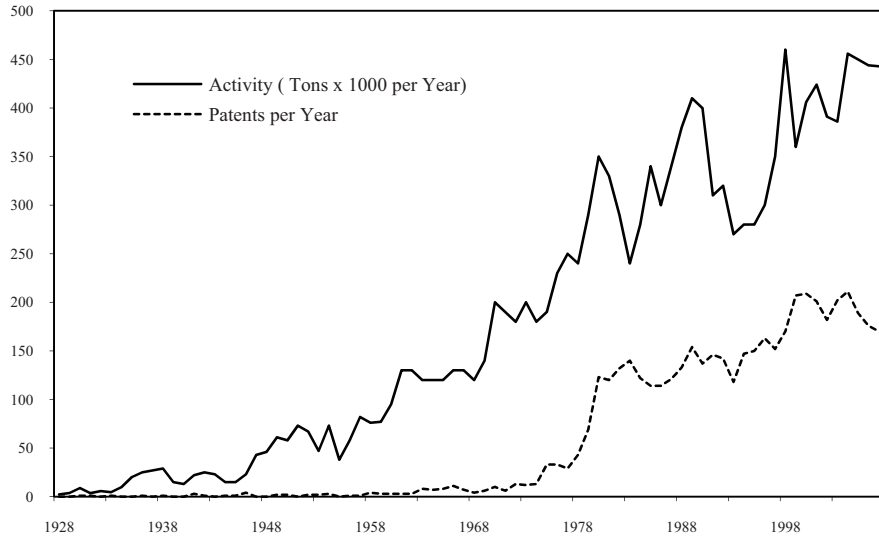


Figure A3.97. Kyanite Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

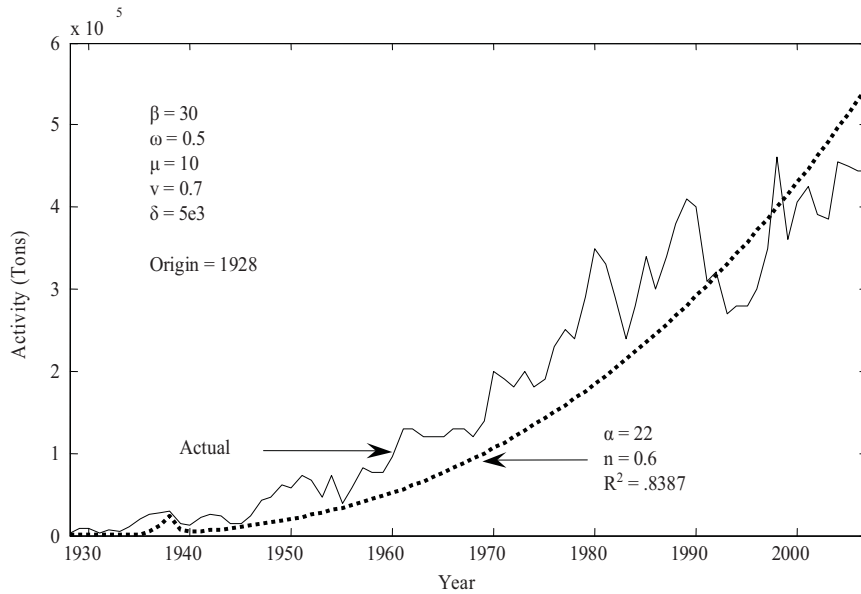


Figure A3.98. USGS World Kyanite Production. World kyanite production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

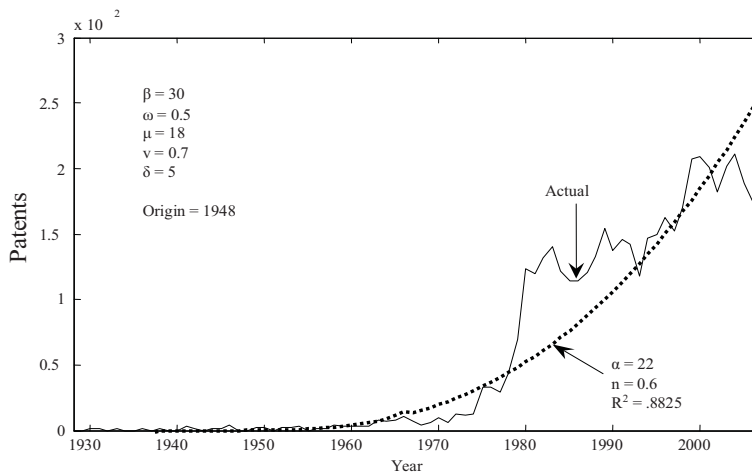


Figure A3.99. EPO Worldwide Patent Search: Kyanite or Aluminum Silicate in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

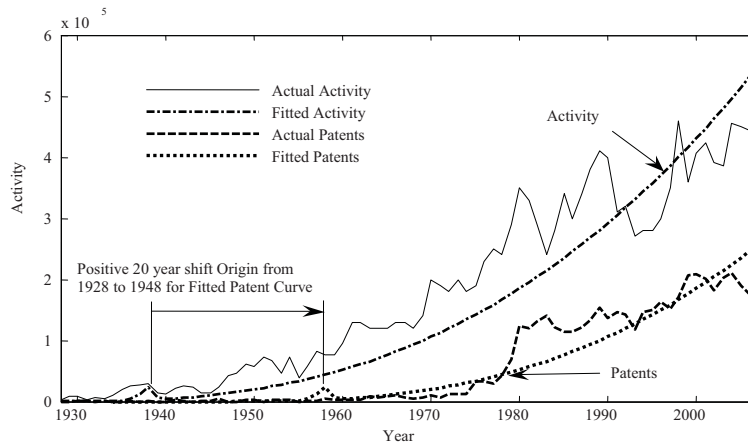


Figure A3.100. Kyanite Best-Fit Activity and Patents. Illustrates best-fit origin shift.

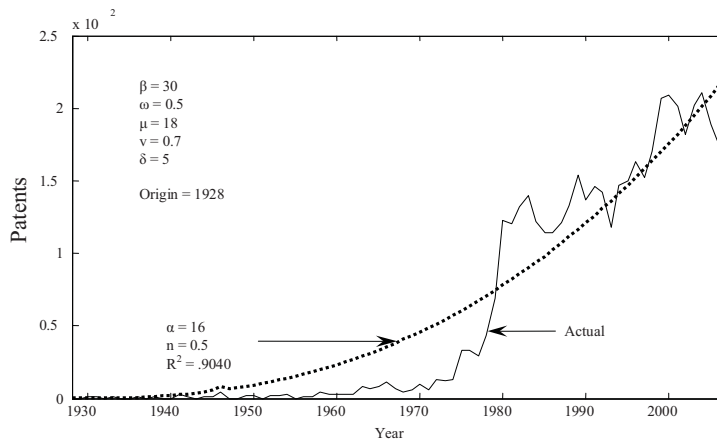


Figure A3.101. Kyanite Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.47 Lead Activity⁵⁸ and Patents⁵⁹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	749000	231	1927	1540000	594	1954	2000000	796	1981	3350000	5204
1901	800000	220	1928	1680000	595	1955	2010000	839	1982	3450000	6114
1902	850000	250	1929	1610000	605	1956	2400000	982	1983	3350000	6262
1903	900000	340	1930	1520000	793	1957	2380000	1093	1984	3200000	7004
1904	950000	307	1931	1260000	792	1958	2350000	919	1985	3390000	7531
1905	1000000	294	1932	1050000	785	1959	2320000	955	1986	3240000	8034
1906	1040000	283	1933	1040000	687	1960	2390000	1342	1987	3430000	7876
1907	993000	334	1934	1200000	679	1961	2390000	1173	1988	3420000	8485
1908	1280000	281	1935	1380000	691	1962	2510000	1174	1989	3400000	10449
1909	1060000	272	1936	1470000	746	1963	2560000	1228	1990	3370000	10534
1910	1100000	297	1937	1590000	720	1964	2530000	1305	1991	3260000	10354
1911	1110000	324	1938	1700000	824	1965	2700000	1535	1992	3200000	11860
1912	1160000	310	1939	1740000	687	1966	2850000	1212	1993	2900000	10598
1913	1150000	342	1940	1700000	631	1967	2870000	1447	1994	2800000	10758
1914	1100000	290	1941	1600000	467	1968	3010000	1396	1995	2710000	9998
1915	1000000	243	1942	1500000	409	1969	3240000	1380	1996	2920000	9867
1916	950000	184	1943	1400000	383	1970	3390000	1580	1997	3100000	9596
1917	900000	178	1944	1300000	314	1971	3490000	1700	1998	3060000	10648
1918	800000	169	1945	1250000	375	1972	3450000	2118	1999	3080000	10770
1919	764000	294	1946	1030000	438	1973	3490000	2027	2000	3200000	11815
1920	804000	323	1947	1310000	575	1974	3490000	1824	2001	3120000	11672
1921	783000	452	1948	1380000	791	1975	3440000	1917	2002	2850000	12153
1922	972000	428	1949	1370000	782	1976	3690000	2179	2003	3150000	11756
1923	1080000	464	1950	1640000	601	1977	3410000	2817	2004	3200000	11619
1924	1220000	489	1951	1600000	663	1978	3460000	2986	2005	3520000	11626
1925	1410000	549	1952	1810000	888	1979	3510000	3674	2006	3650000	11748
1926	1470000	554	1953	1870000	680	1980	3520000	5202	2007	3770000	12083

Table A3.48. Correlation Eq.(A1.1) terms calculated from Table A3.47 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
232875000	339111	6.09E+14	2.86E+09	1.04E+12	1.064E+14	1.79E+09	3.06E+11	0.701272	49.17827

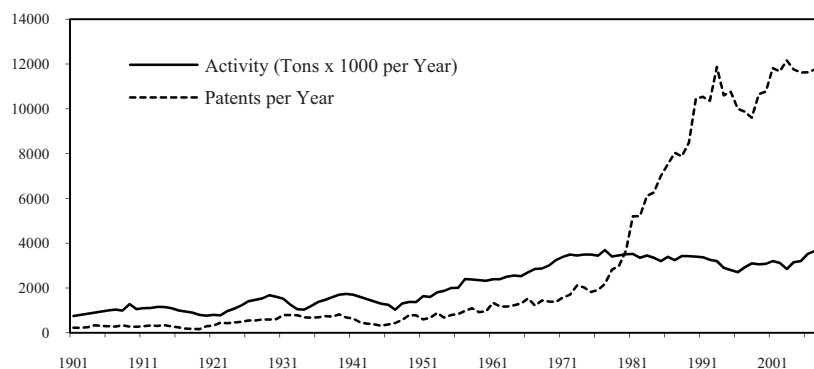


Figure A3.102. Lead Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁵⁸ Activity represents world production of lead, defined at usgs.gov as "...contained lead in world smelter production for the years 1900–54 and for world mine production for the years 1955–98. Data were from the MYB and MR for the years 1900–73 and the MCS for the years 1974 to the most recent. World production data were for contained lead in world smelter production originating from ores and may include secondary lead when inseparable. Blank cells in the worksheet indicate that data were not available for the years 1901–05, 1914–18, 1926, 1937, and 1940–44." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁵⁹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Lead or Pb were used as keywords found in the patent title or abstract by year of publication.

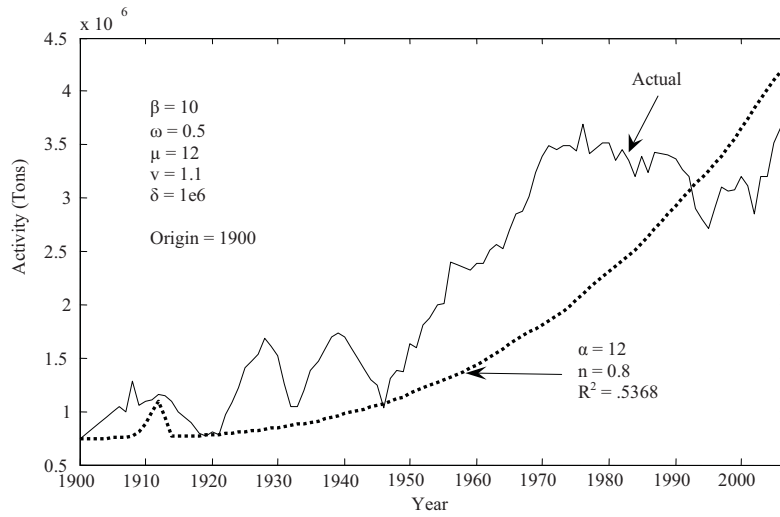


Figure A3.103. USGS World Lead Production. World lead production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

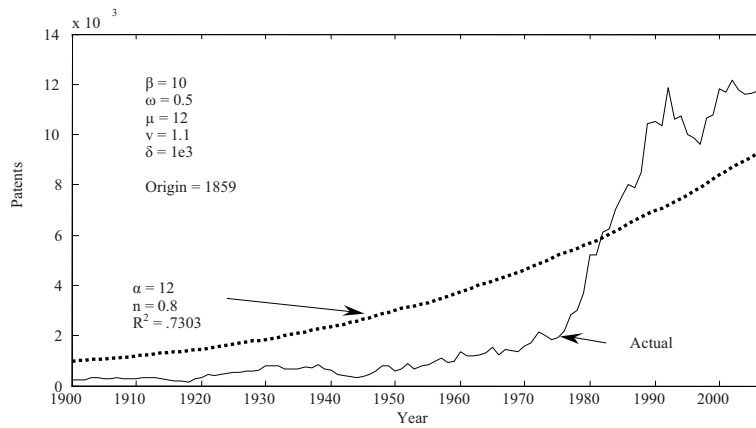


Figure A3.104. EPO Worldwide Patent Search: Lead or Pb in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

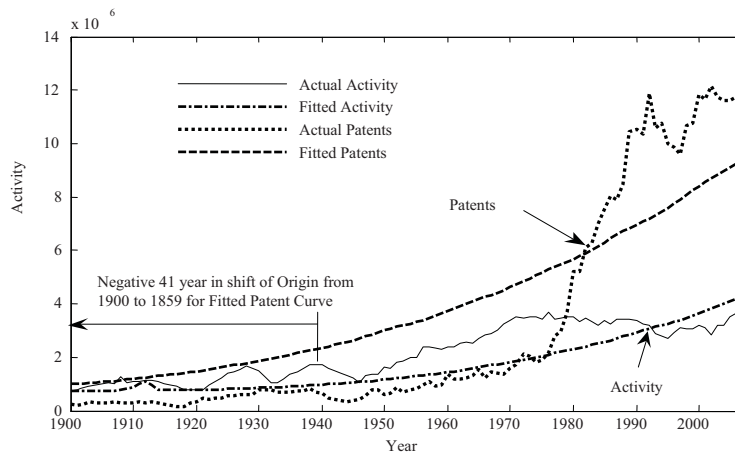


Figure A3.105. Lead Best-Fit Activity and Patents. Illustrates best-fit origin shift.

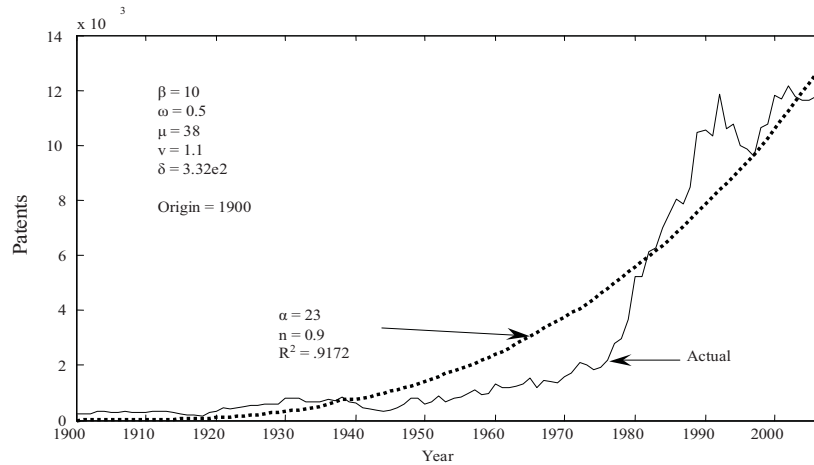


Figure A3.106. Lead Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.49 Lithium Activity⁶⁰ and Patents⁶¹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	5260	15	1954	93200	174	1981	90200	1025
1901			1928	5970	20	1955	86000	193	1982	83600	1230
1902			1929	3140	19	1956	105000	268	1983	93700	1332
1903			1930	3030	25	1957	111000	265	1984	108000	1335
1904			1931	679	50	1958	87800	288	1985	122000	1427
1905			1932	690	40	1959	62400	296	1986	132000	1496
1906			1933	738	50	1960	87100	427	1987	139000	1450
1907			1934	1200	42	1961	57200	392	1988	154000	1499
1908			1935	1540	55	1962	47300	401	1989	173000	1690
1909			1936	2060	34	1963	49500	372	1990	163000	1664
1910			1937	3280	32	1964	64000	418	1991	149000	1660
1911			1938	2510	43	1965	68500	507	1992	156000	1824
1912			1939	3060	51	1966	3450	407	1993	127000	1761
1913			1940	3440	56	1967	7590	482	1994	128000	1954
1914			1941	4400	36	1968	63700	469	1995	177000	2078
1915			1942	6990	17	1969	68000	459	1996	214000	2196
1916			1943	9180	19	1970	73100	553	1997	213000	2504
1917			1944	15600	20	1971	73400	524	1998	178000	3181
1918			1945	2830	28	1972	19700	605	1999	188000	3389
1919			1946	4540	43	1973	79300	521	2000	204000	4013
1920			1947	5350	69	1974	113000	525	2001	210000	4197
1921			1948	4540	99	1975	122000	570	2002	219000	4578
1922			1949	6270	77	1976	75000	655	2003	252000	4450
1923			1950	18000	92	1977	74300	611	2004	262000	4468
1924			1951	25200	94	1978	81900	687	2005	345000	4514
1925	3730	13	1952	25500	158	1979	76000	580	2006	395000	4987
1926	4530	17	1953	57800	155	1980	92800	945	2007	388000	5149

⁶⁰ Activity represents world production of lithium, defined at usgs.gov as "...data are in gross tons of lithium minerals and brine. Since 1967, lithium production was reported as ore and ore concentrates from mines and lithium carbonate from brine deposits. World production data for the years 1966–67 do not include data 4 from Rhodesia (Zimbabwe) and some other African countries. Zimbabwe was by far the largest producer at the time. After 1954, world production does not include U.S. production. Data were not available for the years 1900–24." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁶¹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Lithium or Li were used as keywords found in the patent title or abstract by year of publication.

Table A3.50. Correlation Eq.(A1.1) terms calculated from Table A3.49 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
7236797	85114	1.29E+12	2.43E+08	1.69E+10	6.635E+11	1.56E+08	9.43E+09	0.927197	85.96949

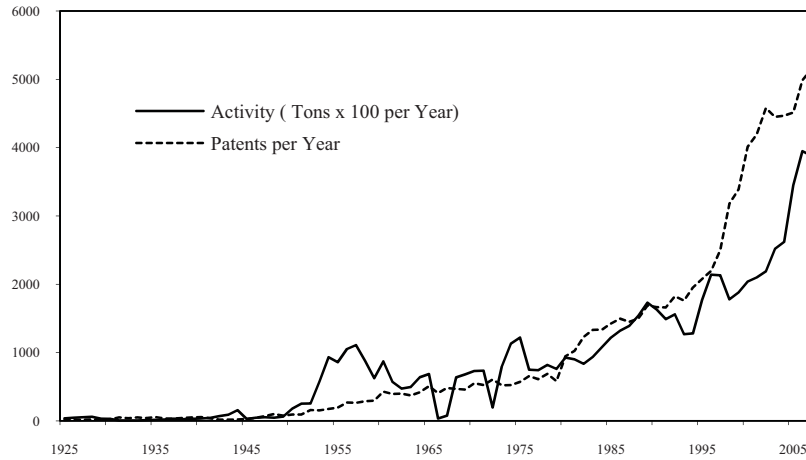


Figure A3.107. Lithium Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

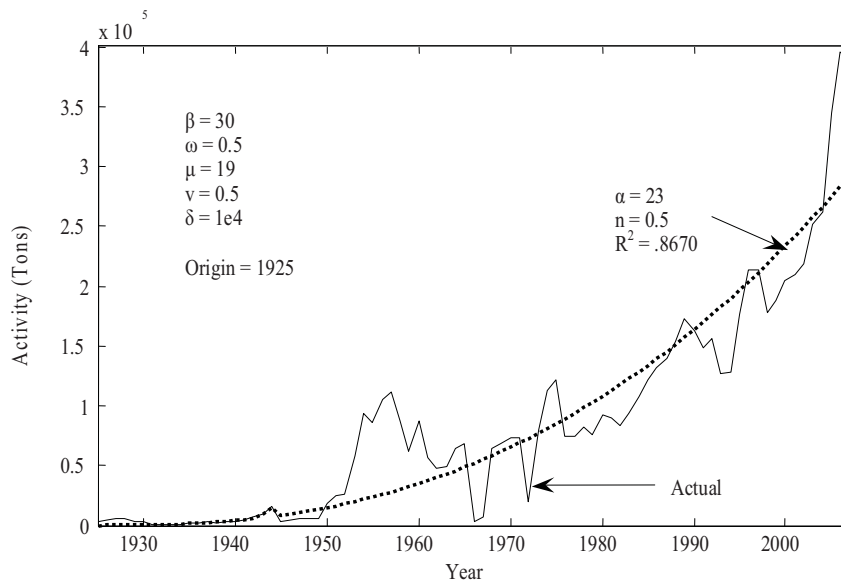


Figure A3.108. USGS World Lithium Production. World lithium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

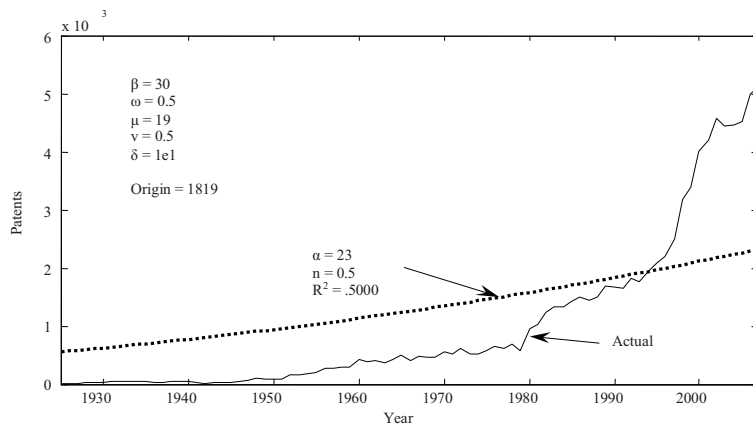


Figure A3.109. EPO Worldwide Patent Search: Lithium or Li in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

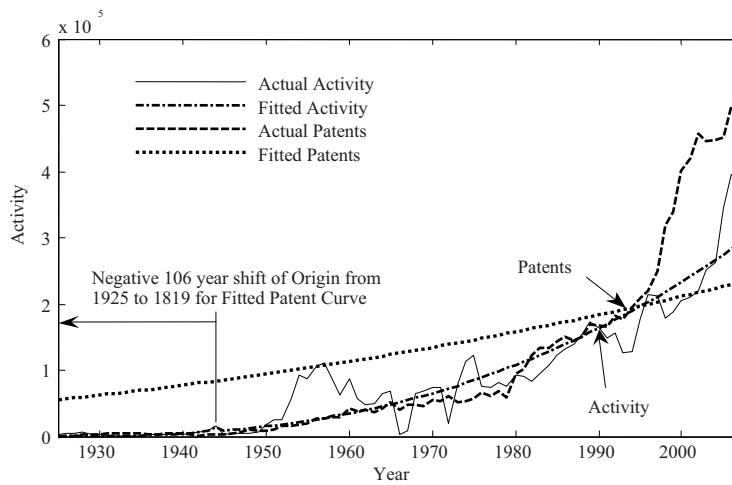


Figure A3.110. Lithium Best-Fit Activity and Patents. Illustrates best-fit origin shift.

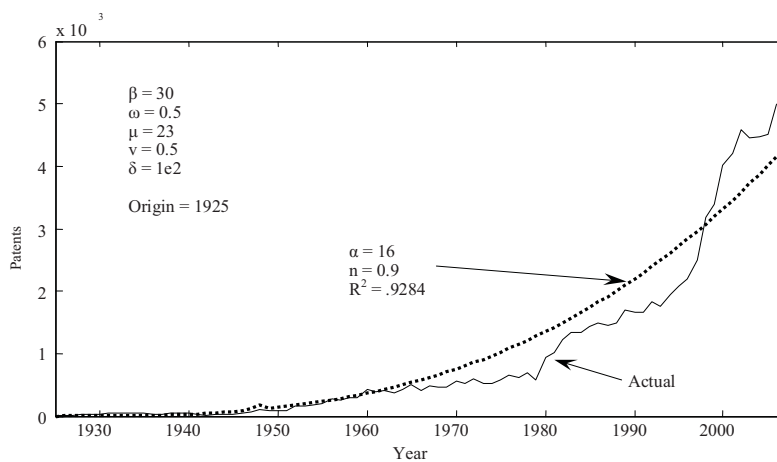


Figure A3.111. Lithium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.51 Magnesite⁶² Activity⁶³ and Patents⁶⁴

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y
1900			1927	833000	29	1954	1650000	86	1981	11300000	86
1901			1928	827000	31	1955	1930000	85	1982	11400000	80
1902			1929	1060000	36	1956	2360000	112	1983	11300000	137
1903			1930	834000	59	1957	2470000	120	1984	11800000	100
1904			1931	691000	73	1958	2330000	122	1985	12200000	105
1905			1932	697000	59	1959	3740000	105	1986	12300000	140
1906			1933	883000	49	1960	6820000	123	1987	12000000	125
1907			1934	1160000	51	1961	7250000	79	1988	12000000	131
1908			1935	1440000	61	1962	7440000	84	1989	12000000	155
1909			1936	1590000	58	1963	8980000	81	1990	10500000	147
1910			1937	2000000	58	1964	9540000	93	1991	9790000	159
1911			1938	1700000	68	1965	10000000	109	1992	10200000	152
1912			1939	2000000	47	1966	10100000	67	1993	8280000	164
1913	556000	11	1940	2000000	52	1967	10200000	91	1994	9020000	151
1914	434000	9	1941	2000000	44	1968	10700000	62	1995	10600000	195
1915	307000	10	1942	2300000	36	1969	9630000	76	1996	11000000	159
1916	599000	8	1943	2400000	30	1970	8720000	87	1997	10100000	160
1917	753000	5	1944	2000000	21	1971	8970000	65	1998	11400000	200
1918	364000	6	1945	1200000	31	1972	8830000	86	1999	9830000	234
1919	284000	9	1946	1200000	41	1973	9070000	77	2000	12700000	227
1920	576000	16	1947	1600000	42	1974	9870000	86	2001	11100000	210
1921	384000	19	1948	2400000	62	1975	9640000	108	2002	14100000	190
1922	536000	25	1949	2700000	59	1976	9070000	115	2003	14100000	237
1923	514000	22	1950	1330000	59	1977	9960000	103	2004	16500000	233
1924	491000	20	1951	1650000	53	1978	10200000	96	2005	15100000	254
1925	716000	20	1952	1520000	70	1979	10900000	70	2006	15000000	216
1926	716000	29	1953	1780000	58	1980	11500000	105	2007	15200000	246

Table A3.52. Correlation Eq.(A1.1) terms calculated from Table A3.51 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
581715000	8632	5.91E+15	1160874	7.7E+10	2.353E+15	376543.2	2.41E+10	0.81084	65.7462

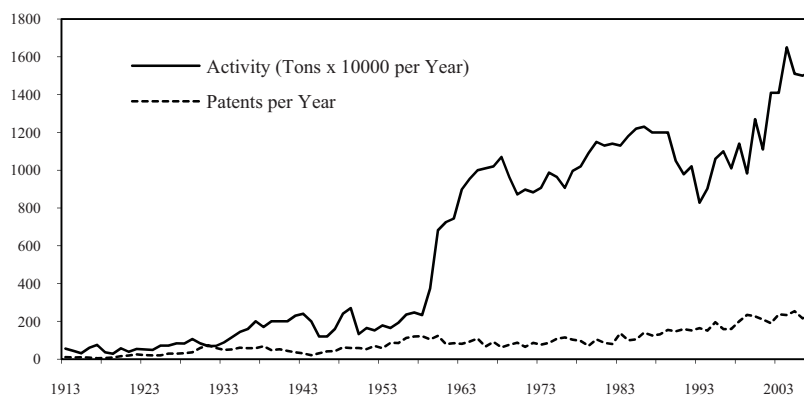


Figure A3.112. Magnesite Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁶² Magnesium carbonate.

⁶³ Activity represents world production of magnesite, defined at usgs.gov as “...metric tons gross weight of magnesite (magnesium carbonate) produced. Data were from the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]. Blank cells for the years 1900–12 in the worksheet indicate that data were not available.” Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁶⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Magnesium and carbonate were used as the keywords found in the patent title or abstract by year of publication.

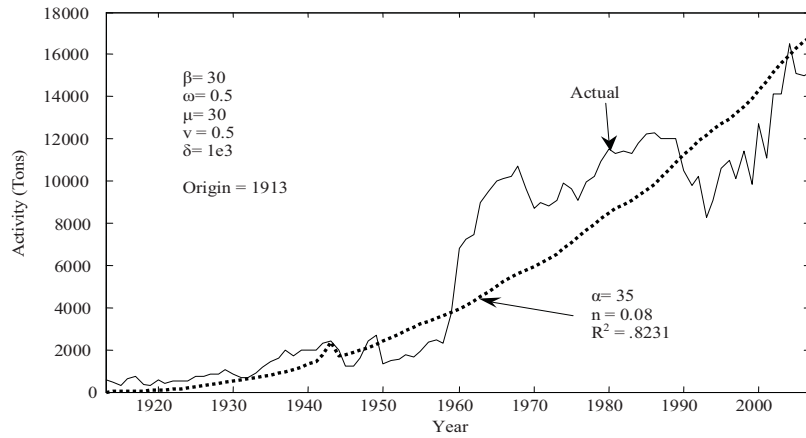


Figure A3.113. USGS World Magnesite Production. World magnesite production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters.

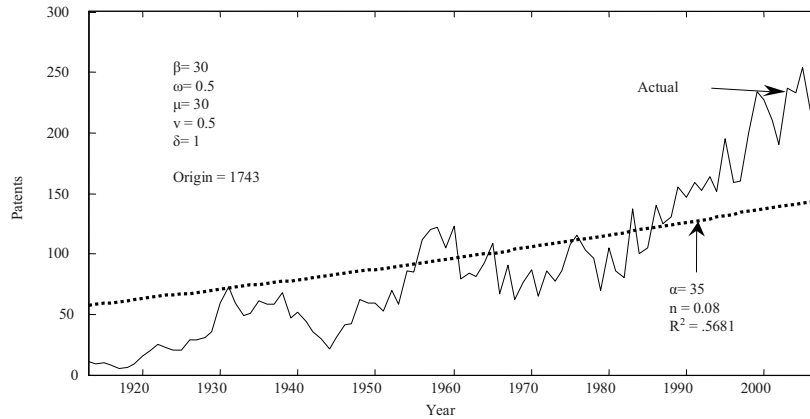


Figure A3.114. EPO Worldwide Patent Search: Magnesium and Carbonate in title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

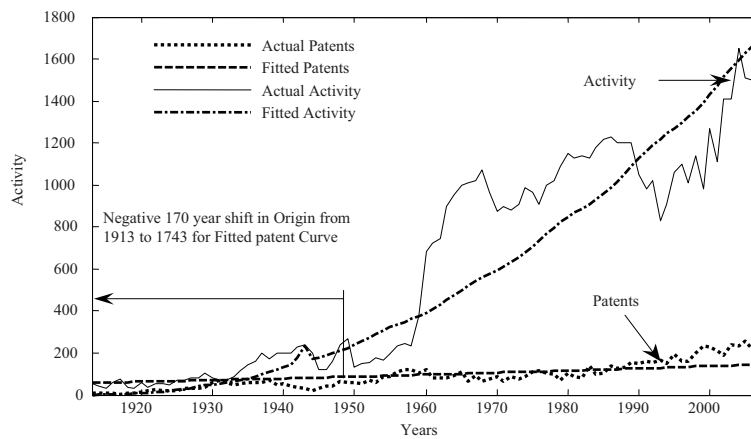


Figure A3.115. Magnesite Best-Fit Activity and Patents. Illustrates best-fit origin shift.

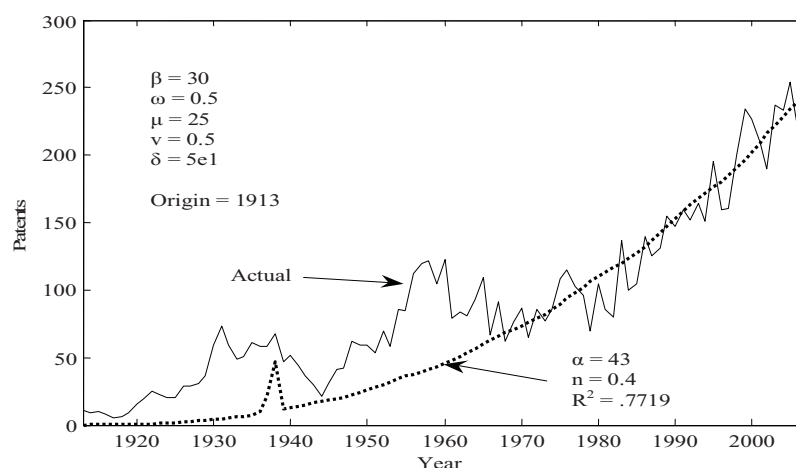


Figure A3.116. Magnesite Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.53 Magnesium Activity⁶⁵ and Patents⁶⁶

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	119000	524	1981	308000	2253
1901			1928			1955	121000	602	1982	254000	2491
1902			1929			1956	103000	697	1983	260000	2604
1903			1930			1957	117000	728	1984	328000	2737
1904			1931			1958	71200	752	1985	325000	2881
1905			1932			1959	74700	731	1986	322000	3224
1906			1933			1960	92900	969	1987	324000	3142
1907			1934			1961	105000	859	1988	334000	3375
1908			1935			1962	134000	787	1989	344000	3832
1909			1936			1963	143000	858	1990	354000	3732
1910			1937	19600	431	1964	151000	947	1991	342000	3975
1911			1938	23900	529	1965	162000	1022	1992	295000	4133
1912			1939	29400	436	1966	163000	901	1993	269000	3901
1913			1940	37800	416	1967	185000	1090	1994	282000	4356
1914			1941	59800	365	1968	189000	1025	1995	395000	4331
1915			1942	105000	308	1969	198000	938	1996	378000	4300
1916			1943	238000	245	1970	220000	1107	1997	384000	4314
1917			1944	218000	232	1971	232000	1103	1998	396000	5485
1918			1945	62000	276	1972	234000	1324	1999	341000	6298
1919			1946	24000	253	1973	240000	1150	2000	422000	7313
1920			1947	32000	326	1974	130000	1009	2001	420000	6851
1921			1948	31000	450	1975	235000	1214	2002	432000	7207
1922			1949	34000	396	1976	249000	1319	2003	509000	7560
1923			1950	45600	330	1977	257000	1398	2004	595000	7770
1924			1951	82600	399	1978	288000	1780	2005	622000	7583
1925			1952	150000	514	1979	307000	1698	2006	689000	7695
1926			1953	153000	408	1980	316000	2339	2007	749000	7980

⁶⁵ Activity represents world production of magnesium, defined at usgs.gov as "...primary magnesium produced. Data were from the MR [Minerals Resources of the United States] and the MYB [Minerals Yearbook]. Blank cells in the worksheet indicate that data were not available for the years 1915–36." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁶⁶ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Magnesium or Mg were used as keywords found in the patent title or abstract by year of publication.

Table A3.54. Correlation Eq.(A1.1) terms calculated from Table A3.53 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
16855500	166508	5.84E+12	7.74E+08	6.36E+10	1.837E+12	3.84E+08	2.41E+10	0.907816	82.41297

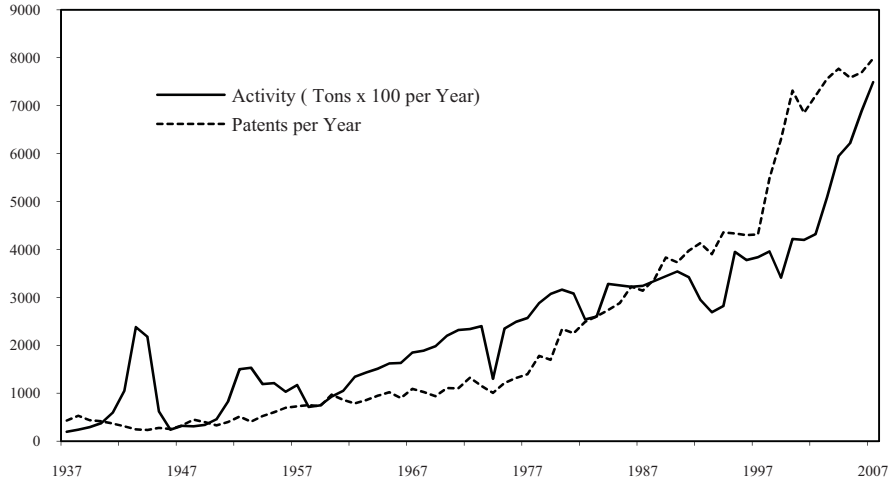


Figure A3.117. Magnesium Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

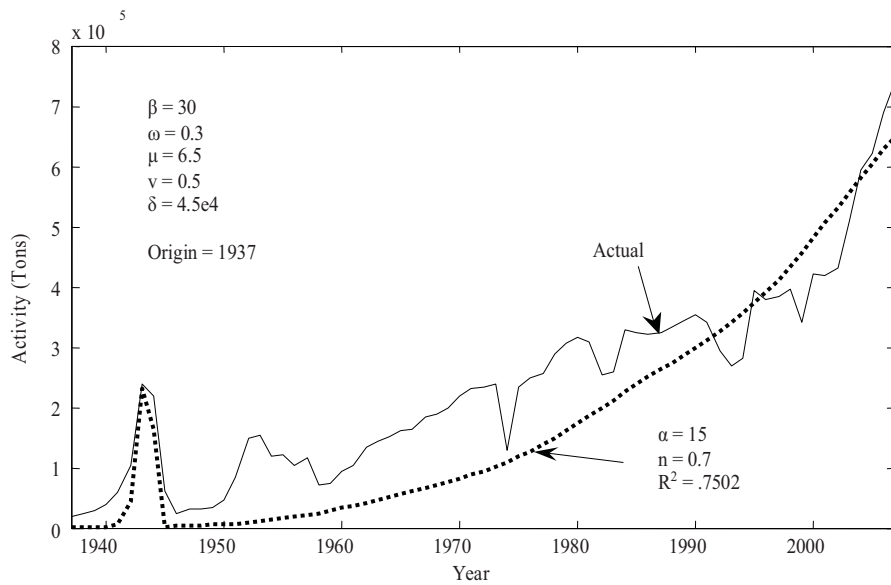


Figure A3.118. USGS World Magnesium Production. World magnesium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

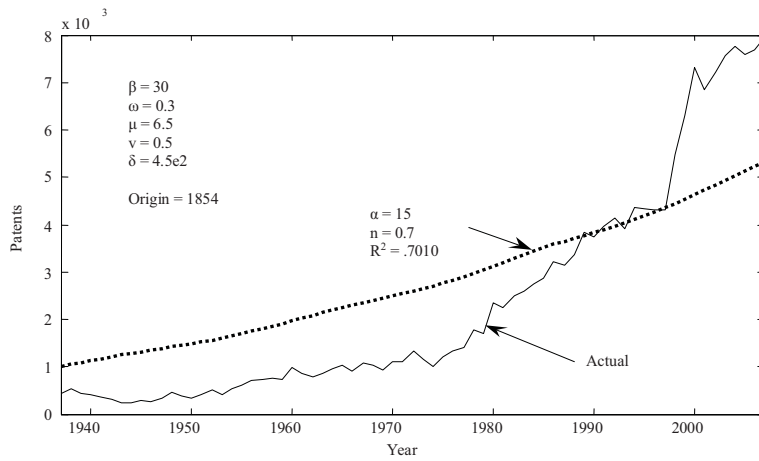


Figure A3.119. EPO Worldwide Patent Search: Magnesium or Mg in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

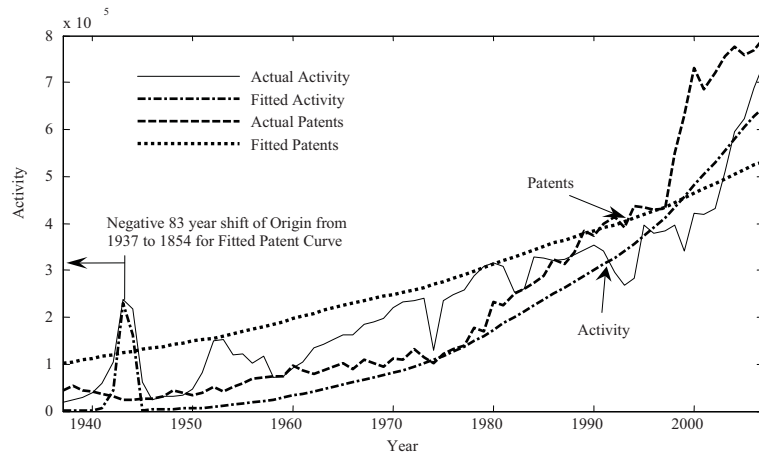


Figure A3.120. Magnesium Best-Fit Activity and Patents. Illustrates best-fit origin shift.

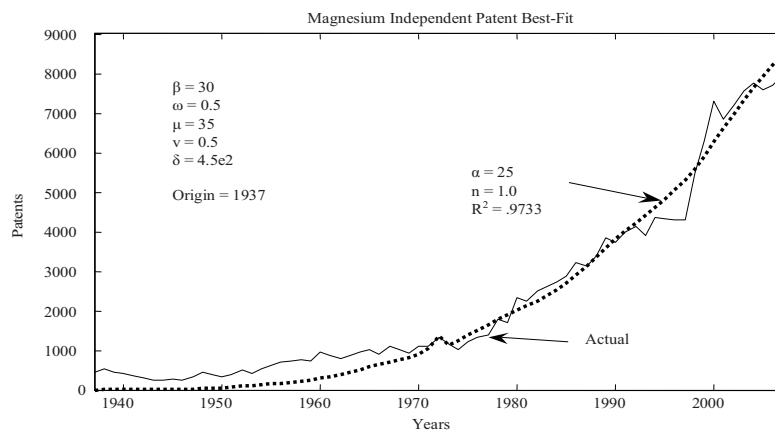


Figure A3.121. Magnesium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.55 Manganese Activity⁶⁷ and Patents⁶⁸

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	592000	38	1927	1430000	116	1954	4500000	261	1981	8400000	2258
1901	429000	25	1928	1280000	156	1955	4870000	310	1982	8580000	2589
1902	441000	45	1929	1580000	200	1956	5310000	410	1983	7780000	2587
1903	411000	51	1930	1590000	199	1957	5820000	406	1984	8600000	2599
1904	416000	49	1931	982000	247	1958	5580000	341	1985	8690000	2851
1905	481000	30	1932	559000	225	1959	5830000	386	1986	8830000	3190
1906	868000	50	1933	779000	206	1960	6120000	549	1987	8340000	3161
1907	1080000	42	1934	1310000	212	1961	6110000	463	1988	8650000	3325
1908	641000	40	1935	1800000	243	1962	6400000	448	1989	9250000	3684
1909	811000	43	1936	2340000	211	1963	6630000	491	1990	9080000	3774
1910	888000	45	1937	2740000	252	1964	7240000	478	1991	7600000	3671
1911	719000	48	1938	2380000	270	1965	7980000	569	1992	7260000	4360
1912	856000	53	1939	1110000	239	1966	8150000	520	1993	7070000	4245
1913	1040000	58	1940	2540000	174	1967	7510000	649	1994	6530000	4308
1914	840000	56	1941	2450000	168	1968	7800000	558	1995	7970000	4425
1915	636000	41	1942	2290000	114	1969	8420000	517	1996	8180000	4390
1916	850000	24	1943	1800000	112	1970	8200000	608	1997	7520000	4454
1917	864000	18	1944	1270000	102	1971	9070000	656	1998	7330000	4862
1918	934000	28	1945	1900000	107	1972	9080000	804	1999	6390000	4836
1919	550000	56	1946	1650000	162	1973	9740000	739	2000	6960000	5767
1920	754000	65	1947	1750000	163	1974	9270000	697	2001	7580000	5658
1921	523000	90	1948	1830000	223	1975	9810000	783	2002	7800000	6277
1922	535000	87	1949	2160000	229	1976	10000000	893	2003	8790000	5973
1923	731000	94	1950	2530000	161	1977	8690000	1006	2004	9900000	6080
1924	919000	99	1951	3180000	202	1978	8690000	1333	2005	11000000	5517
1925	1170000	98	1952	4440000	284	1979	9800000	1411	2006	12200000	5514
1926	1370000	83	1953	4940000	194	1980	9670000	2051	2007	12600000	6018

Table A3.56. Correlation Eq.(A1.1) terms calculated from Table A3.55 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
510129000	141337	3.8E+15	5.59E+08	1.16E+12	1.394E+15	3.74E+08	4.94E+11	0.683516	46.71938

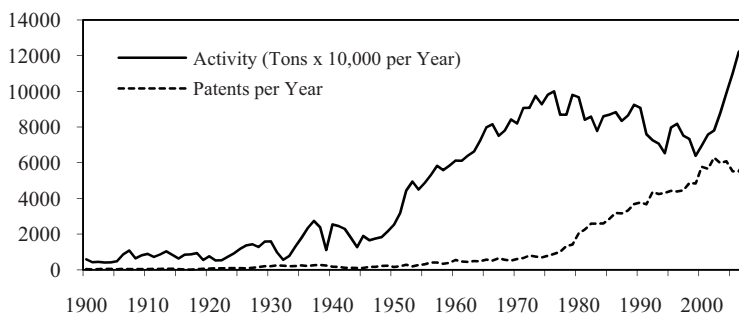


Figure A3.122. Manganese Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁶⁷ Activity represents world production of manganese, defined at usgs.gov as "...contained manganese of world manganese mine production. For the period 1900–50, the reported values were calculated by multiplying gross weight (adjusted to metric tons) reported in Materials Survey—Manganese (U.S. Bureau of Mines and U.S. Geological Survey, 1952) by 0.45 (45 percent). The 45-percent-content estimate was used for consistency with what appears to have been used for the oldest years given in Mineral Facts and Problems. For the period 1951–63, the reported values were calculated by multiplying the gross weight (adjusted to metric tons) reported in MYB series by 0.45. For the period 1964–79, the reported values were taken directly from the values (adjusted to metric tons) reported in the MFP. For the years 1980 to the most recent, the reported values were taken directly from the values published in MYB world production table data series. In more recent years, the manganese content figure has been closer to 35 percent." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁶⁸ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Manganese or Mn were used as keywords found in the patent title or abstract by year of publication.

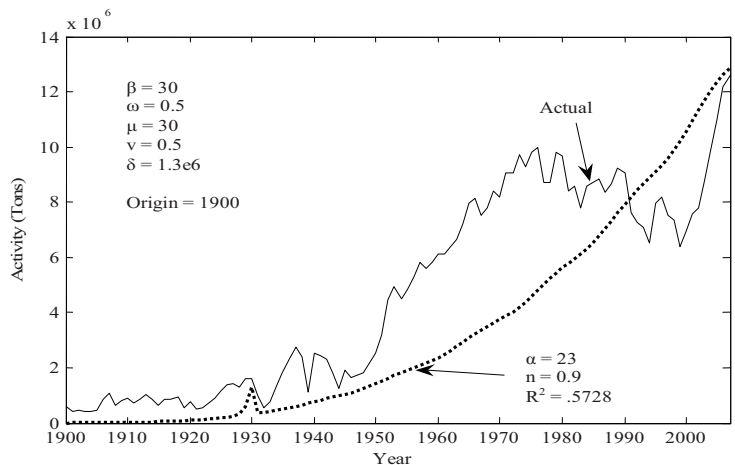


Figure A3.123. USGS World Manganese production. World manganese production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

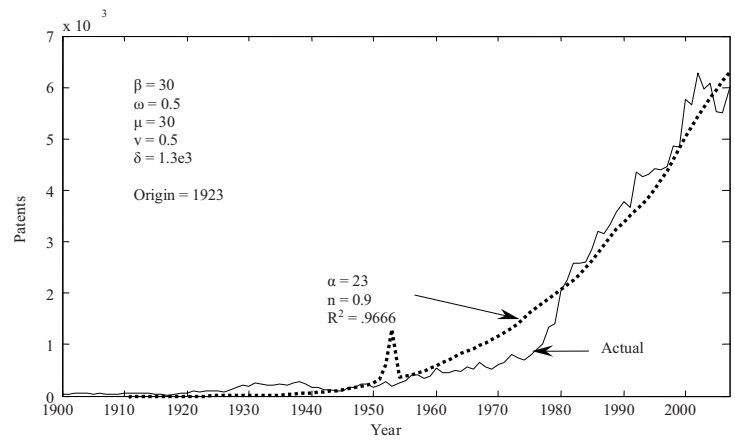


Figure A3.124. EPO Worldwide Patent Search: Manganese or Mn in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

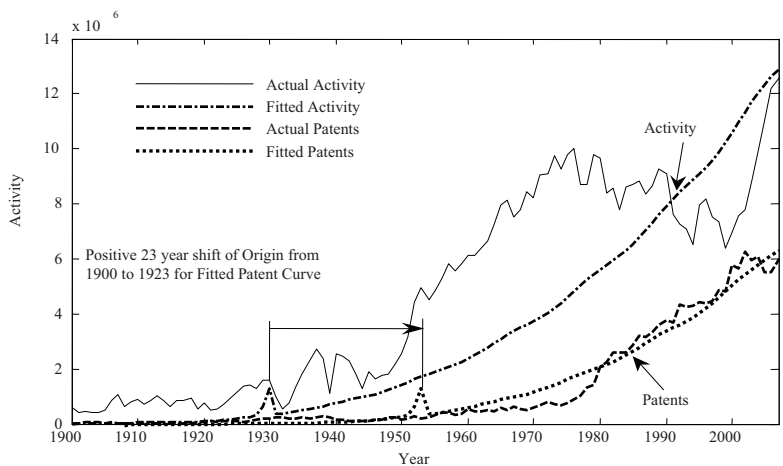


Figure A3.125. Manganese Best-Fit Activity and Patents. Illustrates best-fit origin shift.

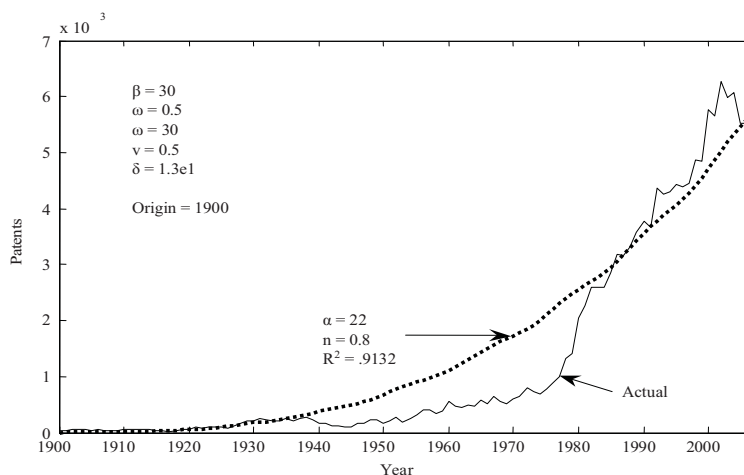


Figure A3.126. Manganese Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.57 Mercury Activity⁶⁹ and Patents⁷⁰

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y
1900	3150	82	1927	5170	216	1954	6180	277	1981	7270	757
1901	2960	80	1928	5140	212	1955	6380	310	1982	6820	896
1902	3800	61	1929	5610	256	1956	7620	350	1983	6230	919
1903	3640	104	1930	3760	333	1957	8260	342	1984	6740	921
1904	3350	140	1931	3420	396	1958	8490	285	1985	6140	1037
1905	3340	151	1932	2850	383	1959	7710	328	1986	7780	1020
1906	3860	147	1933	2060	349	1960	8340	429	1987	5530	890
1907	3310	118	1934	2650	371	1961	8260	380	1988	6840	904
1908	3300	125	1935	3460	366	1962	8430	364	1989	6750	953
1909	3230	112	1936	4270	343	1963	8260	401	1990	4100	893
1910	3690	129	1937	4590	362	1964	8800	445	1991	2540	894
1911	4250	137	1938	5170	463	1965	9230	458	1992	1960	942
1912	4160	136	1939	4830	299	1966	9140	439	1993	1730	922
1913	4050	142	1940	7130	280	1967	8000	516	1994	1960	911
1914	3740	110	1941	9170	194	1968	8950	502	1995	3190	880
1915	3890	100	1942	8990	165	1969	9970	447	1996	2560	797
1916	3500	67	1943	7870	138	1970	9790	505	1997	2410	887
1917	3970	51	1944	5330	108	1971	10400	475	1998	1580	1009
1918	3420	50	1945	4180	131	1972	9620	571	1999	1320	955
1919	3100	80	1946	4960	134	1973	9250	552	2000	1360	1000
1920	2910	118	1947	5360	184	1974	8880	526	2001	1500	901
1921	2130	140	1948	3260	276	1975	8700	554	2002	1490	1061
1922	3170	156	1949	4160	255	1976	8090	583	2003	1730	1068
1923	3210	161	1950	4940	192	1977	6580	590	2004	1900	1076
1924	3070	134	1951	5060	232	1978	6250	642	2005	1520	1111
1925	3560	202	1952	5190	296	1979	6010	554	2006	1150	1151
1926	4000	190	1953	5510	243	1980	6810	756	2007	1170	1157

⁶⁹ Activity represents world production of mercury, defined at usgs.gov as "...from the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]. Mercury production data for the United States, primarily as a byproduct of gold, copper, and zinc mining, were withheld from world mine production data for the years 1993–97 and were not available for the years 1998 to the most recent." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁷⁰ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Mercury or Hg were used as the keywords found in the patent title or abstract by year of publication.

Table A3.58. Correlation Eq.(A1.1) terms calculated from Table A3.51 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
545520	48893	3.43E+09	33904655	2.3702068E+08	672858667	11770160	-9943296	-0.11173	1.248402059

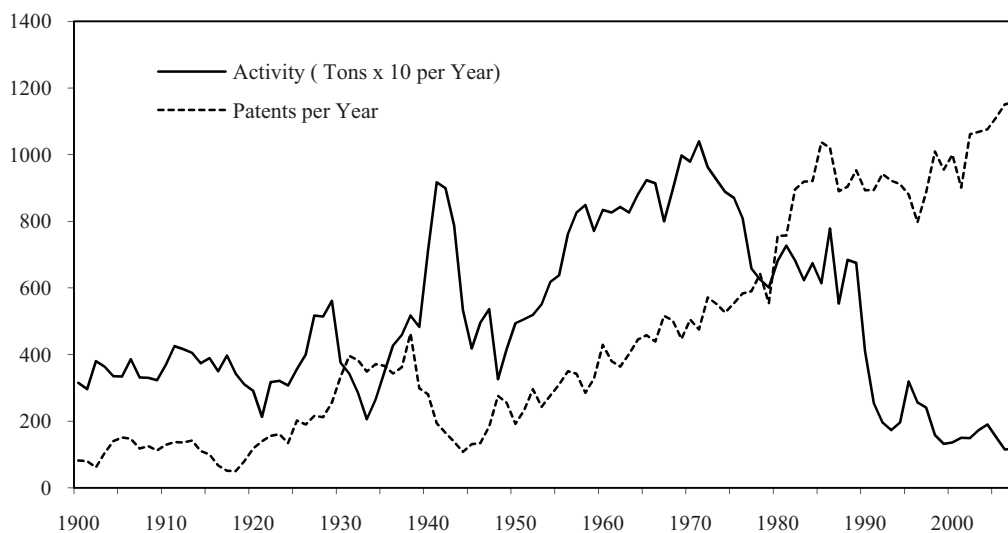


Figure A3.127. Mercury Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

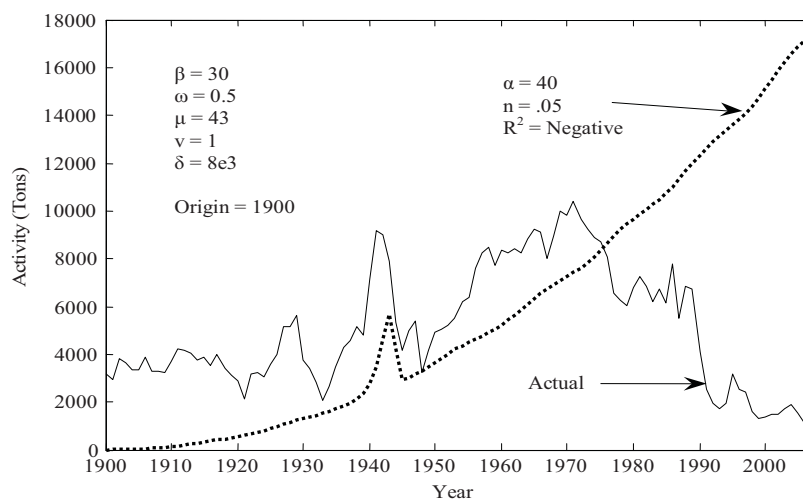


Figure A3.128. USGS World Mercury Production. World mercury production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. Negative R^2 as well as non-linear α indicate possible Stage IV. No best-fit for the patent data was obtainable. Lack of correlation and Stage IV make meaningful activity best-fit impossible.

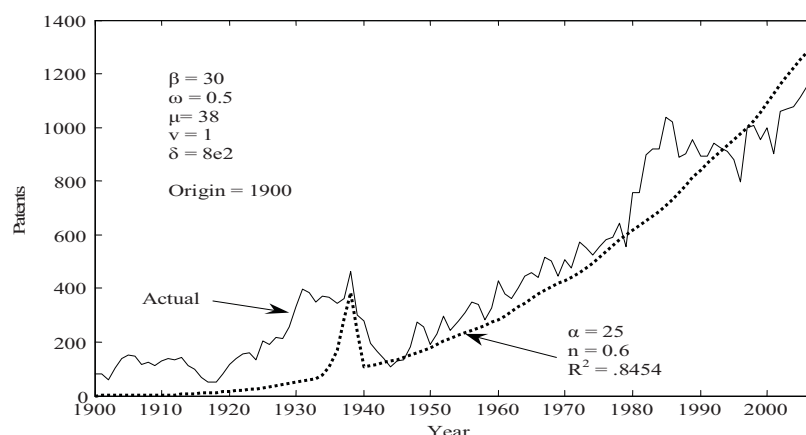


Figure A3.129. Mercury Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.59 Molybdenum Activity⁷¹ and Patents⁷²

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y
1900	10	4	1927	1220	72	1954	29700	206	1981	109000	1675
1901	22	1	1928	1720	109	1955	34000	281	1982	95000	1978
1902	46	2	1929	2000	93	1956	31900	279	1983	63800	2046
1903	148	4	1930	1910	163	1957	34600	337	1984	97700	2039
1904	62	4	1931	1590	197	1958	26200	339	1985	98400	2330
1905	91	8	1932	1320	177	1959	32400	338	1986	93200	2483
1906	91	12	1933	2990	152	1960	40400	464	1987	99500	2574
1907	91	25	1934	5130	139	1961	33600	394	1988	113000	2454
1908	136	20	1935	6530	149	1962	26900	445	1989	136000	2786
1909	91	8	1936	9030	140	1963	34000	524	1990	127000	2631
1910	91	18	1937	14800	173	1964	35300	509	1991	115000	2496
1911	91	25	1938	16400	207	1965	44700	610	1992	114000	2764
1912	181	16	1939	15600	202	1966	56700	568	1993	99200	2734
1913	91	23	1940	17400	114	1967	64300	658	1994	108000	2898
1914	136	18	1941	20300	107	1968	65700	652	1995	136000	2897
1915	272	11	1942	29000	72	1969	72300	668	1996	127000	2757
1916	454	13	1943	31600	72	1970	82300	723	1997	138000	2683
1917	590	13	1944	21500	64	1971	77600	706	1998	135000	3019
1918	816	14	1945	16300	72	1972	79300	900	1999	129000	2983
1919	408	10	1946	10800	112	1973	81700	838	2000	135000	3541
1920	181	27	1947	14000	109	1974	84200	681	2001	133000	3430
1921	45	31	1948	13600	149	1975	81800	746	2002	122000	3892
1922	45	46	1949	11400	145	1976	88700	763	2003	131000	3748
1923	136	36	1950	14500	110	1977	95100	871	2004	159000	3939
1924	272	52	1951	20300	141	1978	100000	1043	2005	186000	3538
1925	680	75	1952	22600	205	1979	104000	1057	2006	184000	3410
1926	816	52	1953	28400	153	1980	111000	1588	2007	205000	3652

⁷¹ Activity represents world production of molybdenum, defined at usgs.gov as "...world mine production of ores and concentrates. For the years 1900–04, data are from Sutolov (1983, p.251-252)[*Sutolov, Alexander, ed., [1983], Statistical summary 1900-1982, in Internet molybdenum yearbook 1983, v. V of Internet molybdenum encyclopedia: Santiago, Chile, Alexander Sutolov Internet Publications, p. 248-297.*]. For the years 1905 to the most recent, data are from the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁷² Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Molybdenum or Mo were used as the keywords found in the patent title or abstract by year of publication.

Table A3.60. Correlation Eq.(A1.1) terms calculated from Table A3.59 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
5395233	98731	5.77E+11	2.45E+08	1.13E+10	3.078E+11	1.55E+08	6.42E+09	0.928979	86.30013

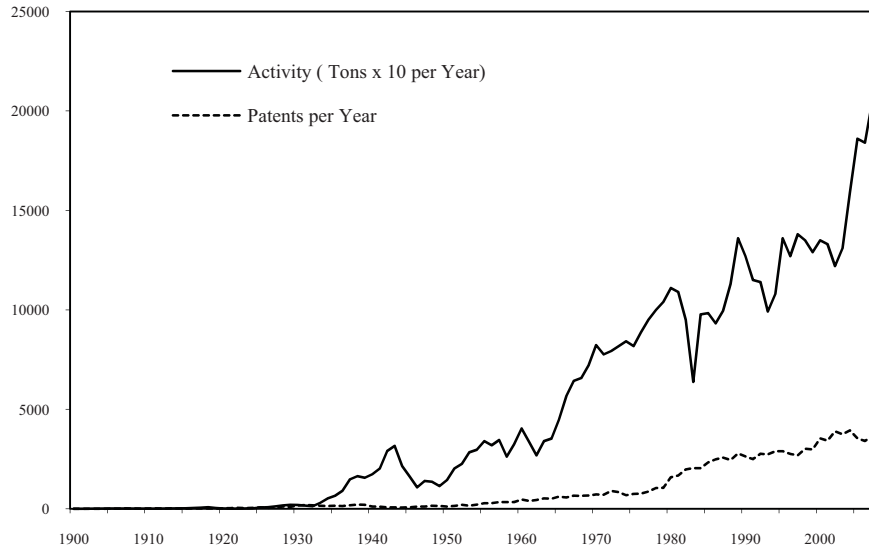


Figure A3.130. Molybdenum Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

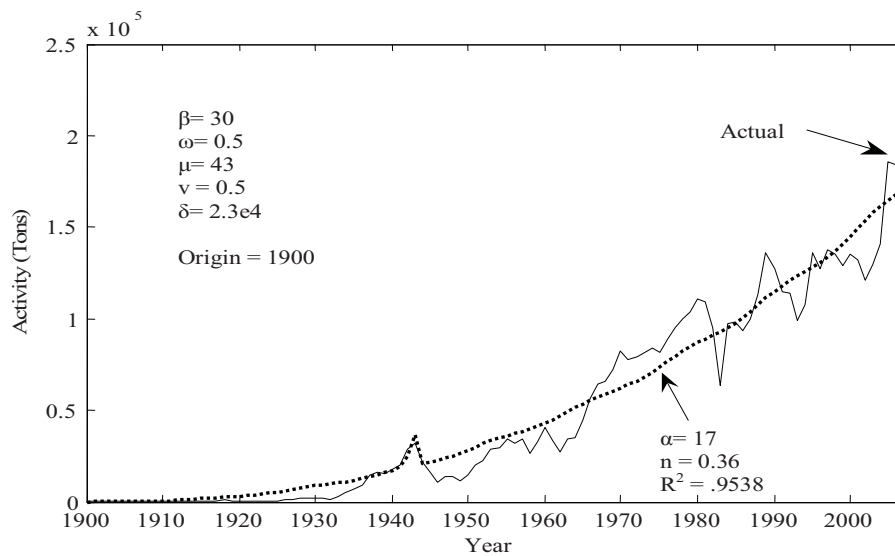


Figure A3.131. USGS World Molybdenum Production. World molybdenum production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

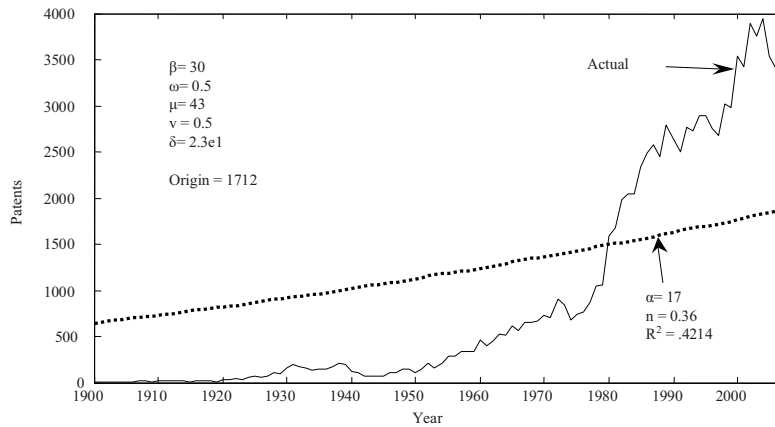


Figure A3.132. EPO Worldwide Patent Search: Molybdenum or Mo in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

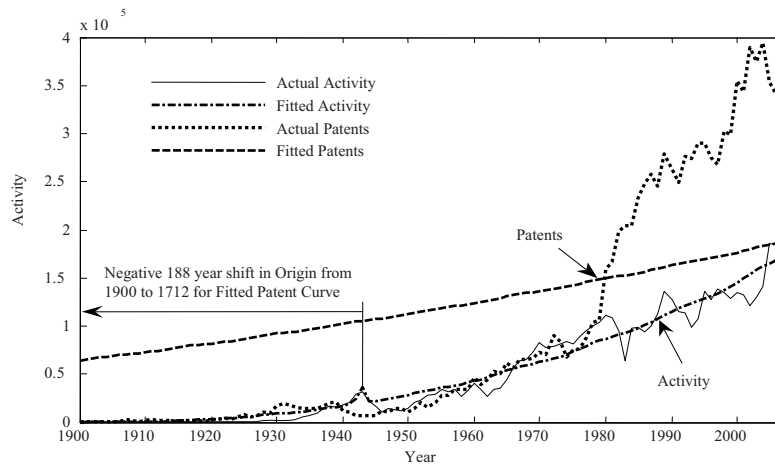


Figure A3.133. Molybdenum Best-Fit Activity and Patents. Illustrates best-fit origin shift.

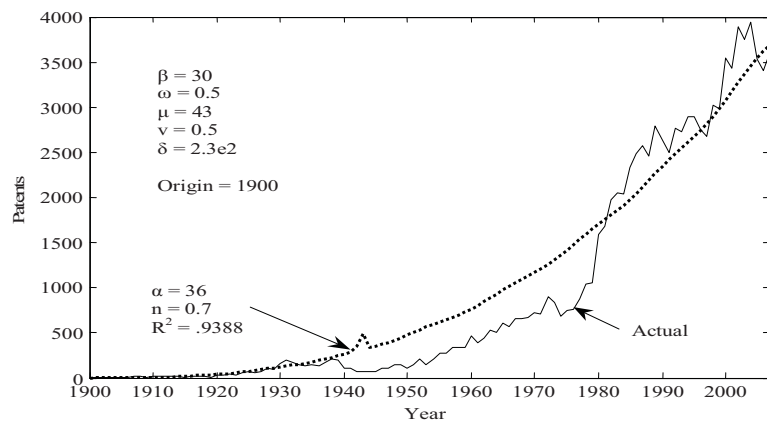


Figure A3.134. Molybdenum Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.61 Nickel Activity⁷³ and Patents⁷⁴

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	9290	37	1927	34500	248	1954	216000	566	1981	726000	3617
1901	11400	37	1928	50300	338	1955	239000	646	1982	621000	4294
1902	12200	48	1929	56300	353	1956	259000	744	1983	673000	4681
1903	10200	54	1930	54200	455	1957	286000	818	1984	773000	4916
1904	10500	56	1931	36300	533	1958	224000	776	1985	813000	5360
1905	15600	57	1932	21800	452	1959	285000	843	1986	852000	5639
1906	16000	54	1933	46300	402	1960	320000	1108	1987	891000	5552
1907	16300	68	1934	71600	388	1961	361000	1091	1988	952000	5850
1908	14900	76	1935	77400	441	1962	357000	1033	1989	987000	6489
1909	17000	79	1936	93400	399	1963	339000	1070	1990	974000	6108
1910	23100	73	1937	120000	434	1964	371000	1160	1991	1010000	5879
1911	25200	80	1938	115000	496	1965	425000	1394	1992	1010000	6657
1912	27900	68	1939	122000	438	1966	412000	1323	1993	928000	6006
1913	32200	98	1940	140000	359	1967	449000	1670	1994	932000	6218
1914	30000	83	1941	162000	290	1968	497000	1560	1995	1040000	5949
1915	39100	75	1942	158000	203	1969	487000	1591	1996	1060000	5906
1916	45500	48	1943	167000	207	1970	628000	1815	1997	1140000	5780
1917	46200	37	1944	157000	167	1971	637000	1841	1998	1180000	6605
1918	47600	53	1945	145000	214	1972	611000	2189	1999	1170000	6597
1919	23100	76	1946	123000	282	1973	710000	1986	2000	1290000	7656
1920	35700	111	1947	140000	290	1974	770000	1807	2001	1350000	7387
1921	10400	161	1948	151000	410	1975	802000	2057	2002	1350000	7636
1922	11800	142	1949	146000	437	1976	792000	2315	2003	1370000	7602
1923	31100	143	1950	145000	360	1977	828000	2402	2004	1420000	7542
1924	35300	130	1951	132000	410	1978	658000	2717	2005	1490000	6815
1925	37100	186	1952	146000	573	1979	686000	2657	2006	1580000	6717
1926	33900	179	1953	198000	454	1980	779000	3660	2007	1660000	7025

Table A3.62. Correlation Eq.(A1.1) terms calculated from Table A3.61 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
46347690	221664	4.19E+13	1.14E+09	2.13E+11	2.198E+13	6.86E+08	1.17E+11	0.956269	91.44502

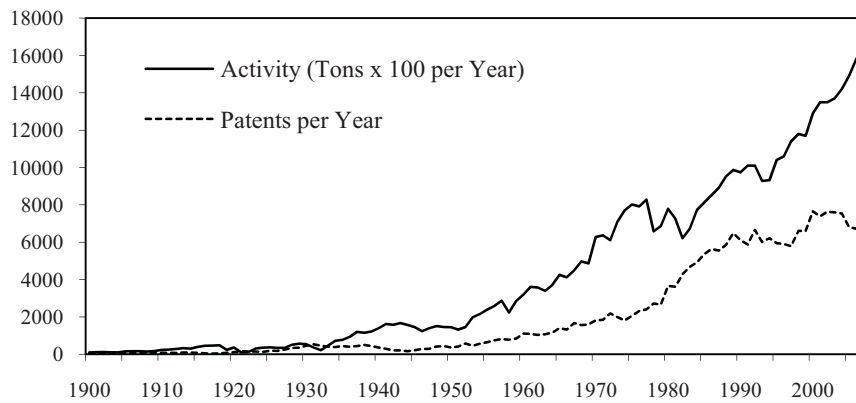


Figure A3.135. Nickel Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁷³ Activity represents world production of nickel, defined at usgs.gov as "...mine production and is reported as recoverable nickel contained in the ore mined. Where actual mine output was not available, data related to a more highly processed form were used to indicate the minimum magnitude of mine output. In 1953, production data for countries once comprising the former Soviet Union were included for the first time. Data are sourced as follows: 1900–29, MS50 [*Materials Survey- Nickel 1950*]; and 1930 to the most recent year, MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁷⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Nickel or Ni were used as keywords found in the patent title or abstract by year of publication.

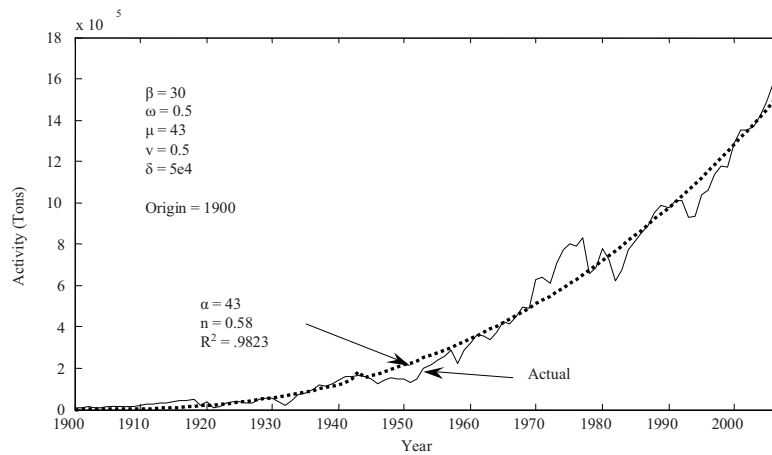


Figure A3.136. USGS World Nickel Production. World nickel production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

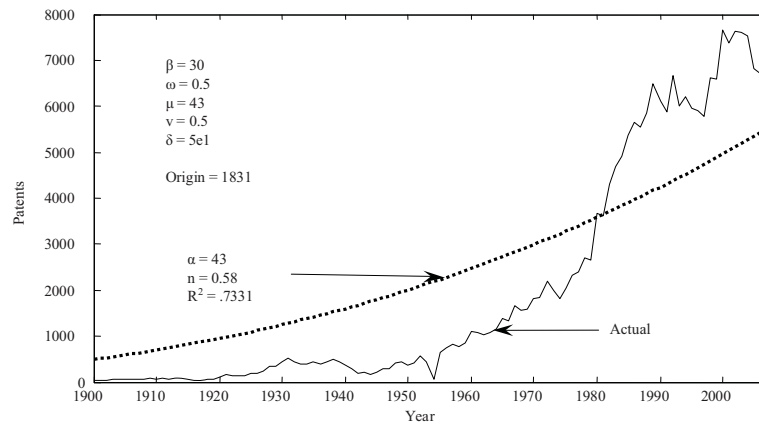


Figure A3.137. EPO Worldwide Patent search: Nickel or Ni in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

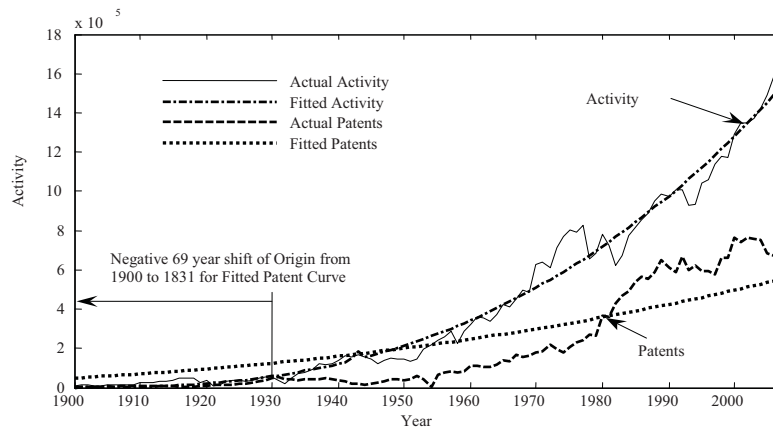


Figure A3.138. Nickel Best-Fit Activity and Patents. Illustrates best-fit origin shift.

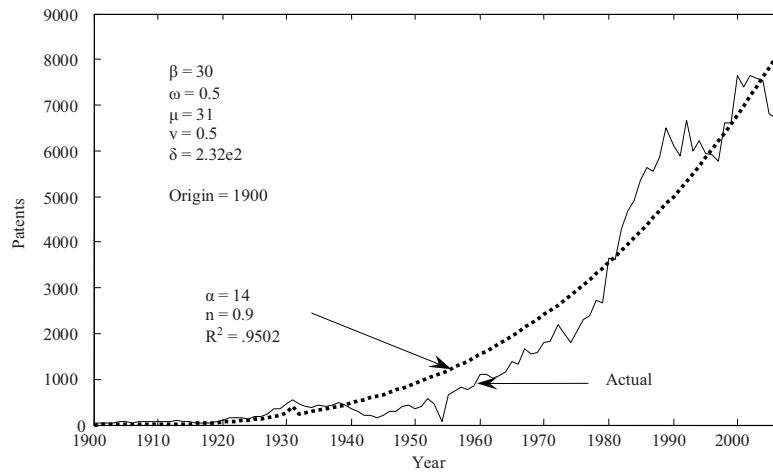


Figure A3.139. Nickel Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.63 Niobium Activity⁷⁵ and Patents⁷⁶

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981	14800	671
1901			1928			1955			1982	10600	873
1902			1929			1956			1983	8580	1034
1903			1930			1957			1984	13900	993
1904			1931			1958			1985	14800	1107
1905			1932			1959			1986	14600	1282
1906			1933			1960			1987	9360	1325
1907			1934			1961			1988	16900	1453
1908			1935			1962			1989	14100	1591
1909			1936			1963			1990	15300	1567
1910			1937			1964	2480	181	1991	15700	1484
1911			1938			1965	3120	282	1992	15300	1828
1912			1939			1966	5060	275	1993	12400	1905
1913			1940			1967	5150	296	1994	15700	1934
1914			1941			1968	4950	270	1995	15600	1819
1915			1942			1969	6610	302	1996	16200	1711
1916			1943			1970	8460	314	1997	20500	1656
1917			1944			1971	3740	342	1998	26200	1797
1918			1945			1972	5950	380	1999	24600	1729
1919			1946			1973	14700	308	2000	24800	1981
1920			1947			1974	9340	292	2001	31100	1928
1921			1948			1975	7860	361	2002	33000	2423
1922			1949			1976	9470	335	2003	40400	2286
1923			1950			1977	8800	408	2004	41900	2385
1924			1951			1978	9670	445	2005	60300	2021
1925			1952			1979	14000	448	2006	51200	2095
1926			1953			1980	15100	645	2007	60400	2168

⁷⁵ Activity represents world production of niobium, defined at usgs.gov as “...the niobium content of niobium-bearing ores and mineral concentrates that were produced from mines throughout the world. Data for the years 1964–68 were recorded from the MFP [*Mineral Facts and Problems*]and the MCP [*Mineral Commodity Summaries*], and for the years 1969 to the most recent were recorded from the MYB [*Minerals Yearbook*].” Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁷⁶ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Niobium or Nb or Columbian were used as keywords found in the patent title or abstract by year of publication.

Table A3.64. Correlation Eq.(A1.1) terms calculated from Table A3.63 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
772700	50930	2.22E+10	82904308	1.24E+09	8.637E+09	23952833	3.43E+08	0.754531	56.93172

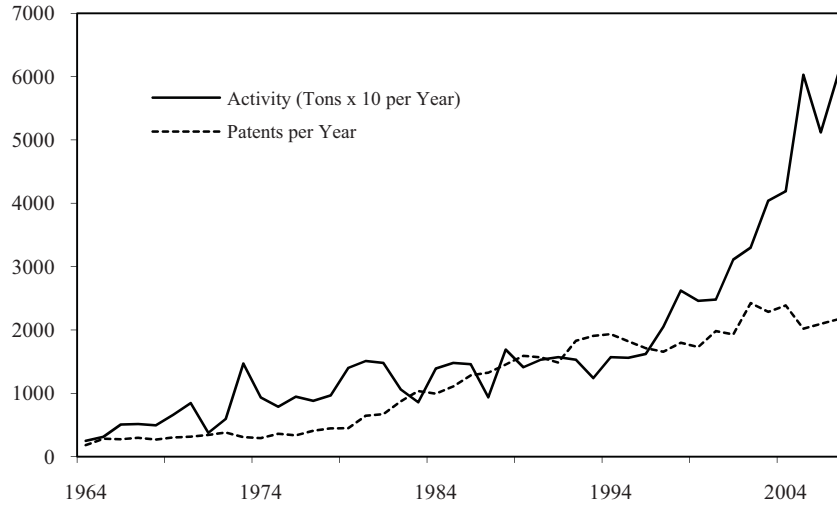


Figure A3.140. Niobium Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

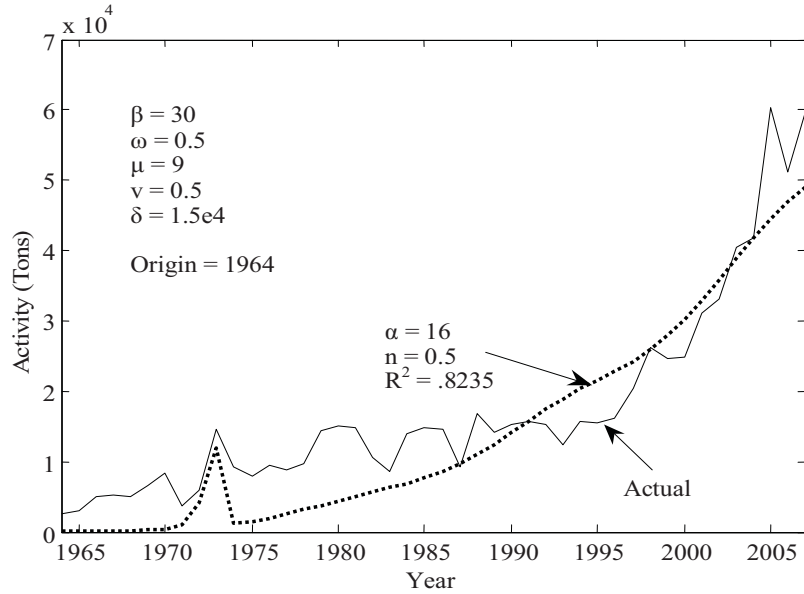


Figure A3.141. USGS World Niobium Production. World niobium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

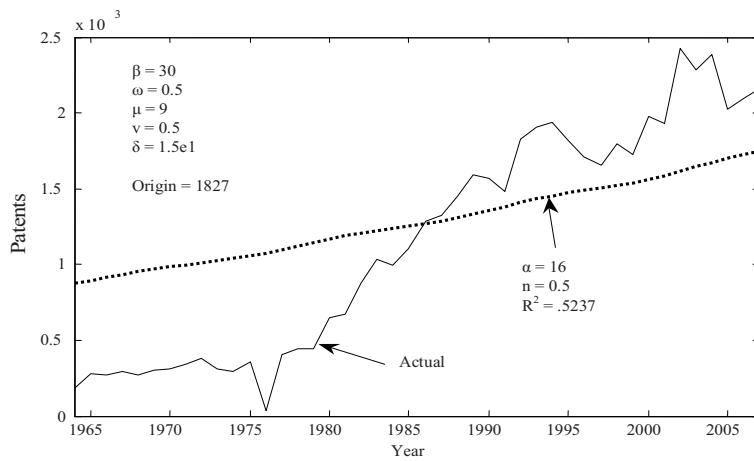


Figure A3.142. EPO Worldwide Patent Search: Niobium, Nb or Columbium in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

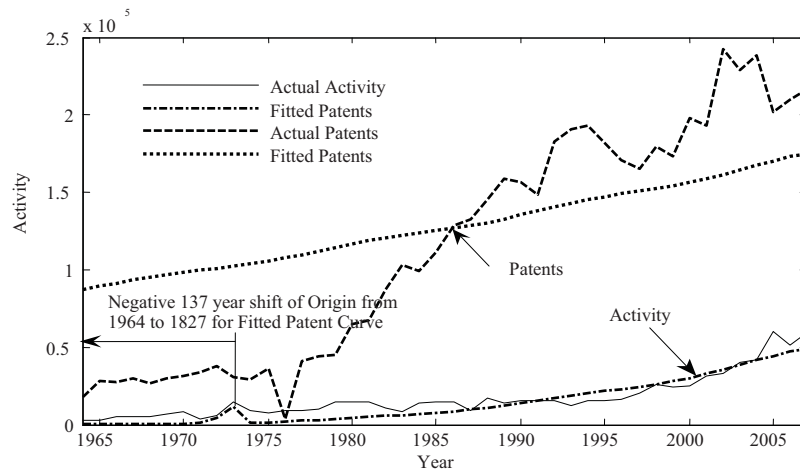


Figure A3.143. Niobium Best-Fit Activity and Patents. Illustrates best-fit origin shift.

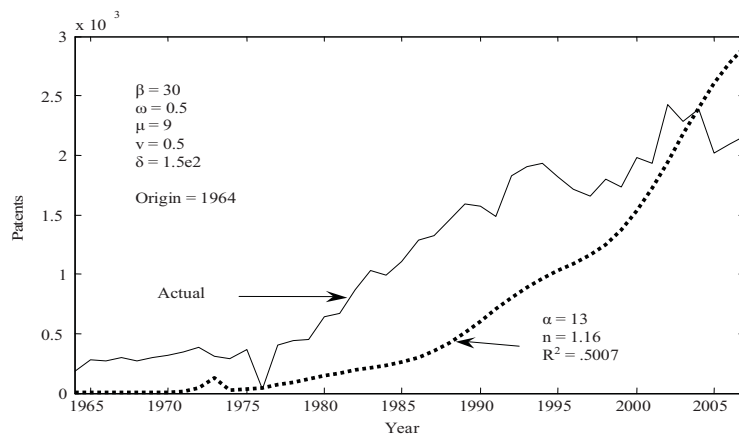


Figure A3.144. Niobium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.65 Nitrogen Activity⁷⁷ and Patents⁷⁸

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y
1900			1927			1954	7300000	387	1981	77000000	2601
1901			1928			1955	8070000	486	1982	75900000	2702
1902			1929			1956	8620000	567	1983	80400000	2746
1903			1930			1957	9270000	668	1984	88600000	2868
1904			1931			1958	10800000	658	1985	91000000	3225
1905			1932			1959	11800000	651	1986	91100000	3718
1906			1933			1960	12900000	920	1987	95100000	3892
1907			1934			1961	14000000	878	1988	99300000	4634
1908			1935			1962	11900000	910	1989	99300000	5372
1909			1936			1963	17100000	968	1990	97500000	5299
1910			1937			1964	19400000	1077	1991	93800000	5314
1911			1938			1965	21800000	1202	1992	93400000	6091
1912			1939			1966	25000000	999	1993	91600000	5700
1913			1940			1967	28700000	1230	1994	93800000	6041
1914			1941			1968	32100000	1163	1995	100000000	6039
1915			1942			1969	35900000	1059	1996	105000000	5676
1916			1943			1970	38800000	1098	1997	103000000	5907
1917			1944			1971	41100000	1023	1998	104000000	6945
1918			1945			1972	43000000	1391	1999	107000000	7085
1919			1946	2380000	175	1973	46700000	1284	2000	108000000	8344
1920			1947	3330000	245	1974	48400000	1170	2001	105000000	8452
1921			1948	3950000	299	1975	49500000	1634	2002	109000000	8885
1922			1949	4560000	308	1976	56900000	1935	2003	110000000	8761
1923			1950	4810000	208	1977	62000000	2111	2004	117000000	9024
1924			1951	5240000	273	1978	67200000	2039	2005	123000000	7868
1925			1952	5300000	382	1979	71100000	1897	2006	126000000	7508
1926			1953	6450000	324	1980	73600000	2442	2007	131000000	7844

Table A3.66. Correlation Eq.(A1.1) terms calculated from Table A3.65 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
3.625E+09	192602	3.18E+17	1.09E+09	1.79E+13	1.063E+17	4.89E+08	6.61E+12	0.916433	83.98487

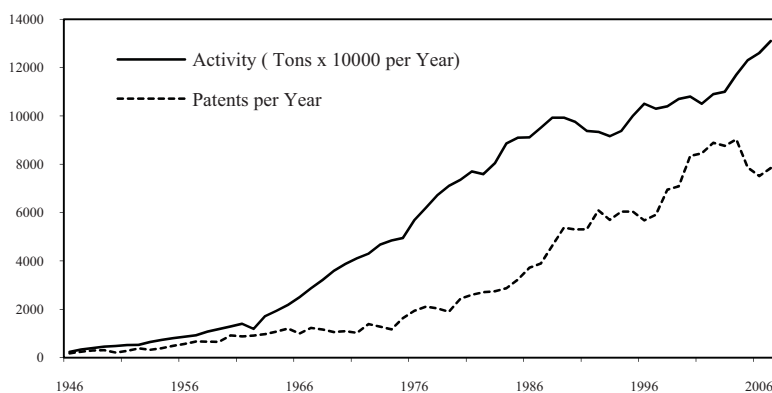


Figure A3.145. Nitrogen Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁷⁷ Activity represents world production of nitrogen, defined at usgs.gov as "... ammonia produced. Data for 1946–57 were for "fertilizer nitrogen compounds," and were reported as fertilizer years (July 1–June 30), not calendar years. Blank cells in the worksheet indicate that data were not available for the years 1943–45. Data were from the MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁷⁸ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Nitrogen was used as the keyword found in the patent title or abstract by year of publication.

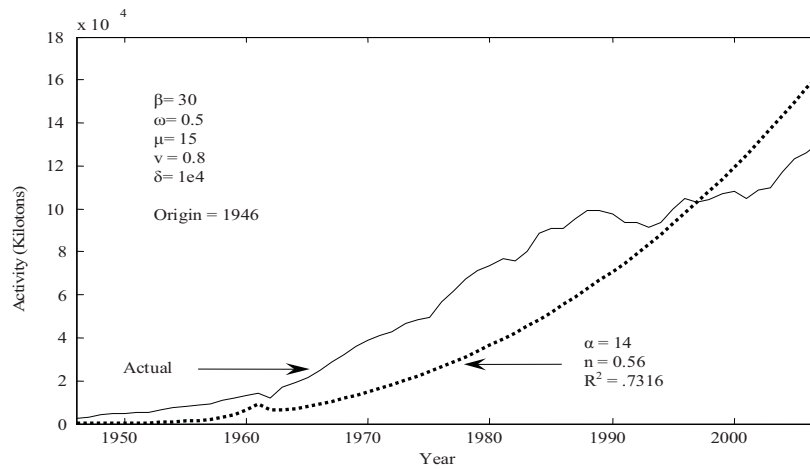


Figure A3.146. USGS World Nitrogen Production. World nitrogen production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

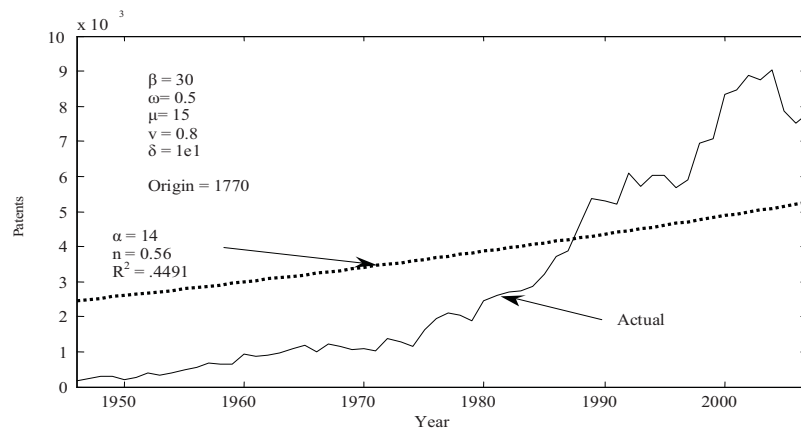


Figure A3.147. EPO Worldwide Patent Search: Nitrogen in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

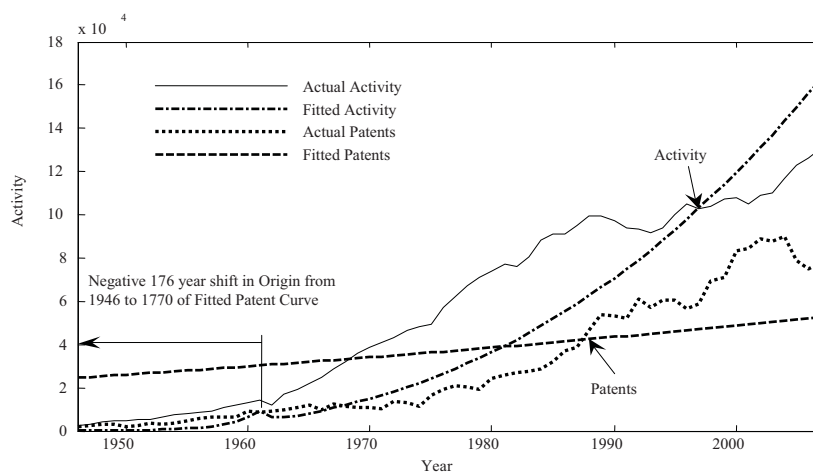


Figure A3.148. Nitrogen Best-Fit Activity and Patents. Illustrates best-fit origin shift.

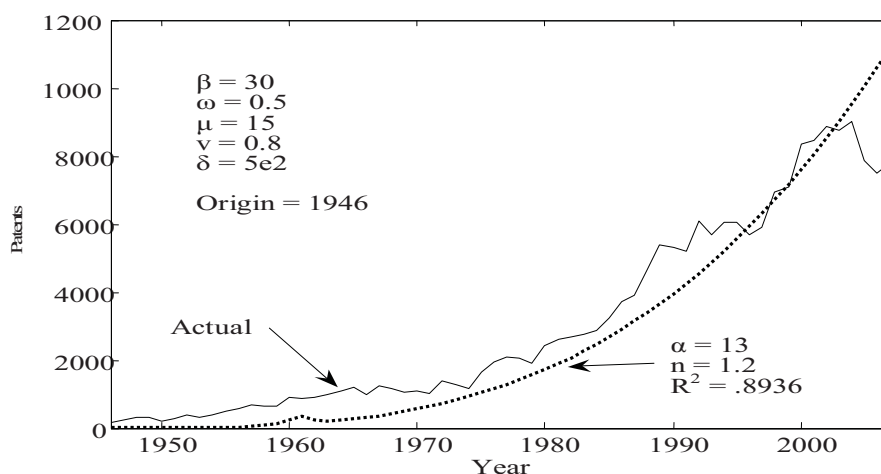


Figure A3.149. Nitrogen Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.67 Phosphate Rock⁷⁹ Activity⁸⁰ and Patents⁸¹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (pat)
1900	3150000	16	1927	9990000	78	1954	30500000	385	1981	145000000	1253
1901	3000000	13	1928	10100000	136	1955	30500000	424	1982	129000000	1370
1902	3120000	13	1929	10400000	147	1956	34200000	420	1983	143000000	1376
1903	3450000	19	1930	11800000	274	1957	33200000	534	1984	154000000	1411
1904	3870000	20	1931	7860000	318	1958	33700000	426	1985	151000000	1532
1905	3850000	19	1932	7110000	260	1959	38400000	531	1986	141000000	1639
1906	4190000	17	1933	8900000	214	1960	41800000	638	1987	147000000	1605
1907	4720000	17	1934	9510000	203	1961	45500000	558	1988	166000000	1614
1908	5380000	13	1935	10500000	227	1962	63300000	570	1989	163000000	1878
1909	4950000	24	1936	11300000	207	1963	54600000	597	1990	162000000	1932
1910	5430000	29	1937	12900000	219	1964	63700000	601	1991	150000000	1920
1911	5940000	28	1938	12900000	241	1965	71400000	701	1992	139000000	2170
1912	6730000	21	1939	12800000	202	1966	84500000	638	1993	119000000	1974
1913	7230000	22	1940	10300000	146	1967	87300000	811	1994	127000000	1931
1914	5420000	25	1941	10800000	147	1968	94100000	683	1995	130000000	2087
1915	4120000	21	1942	8800000	123	1969	92100000	688	1996	135000000	2085
1916	4830000	20	1943	9250000	114	1970	95100000	735	1997	143000000	2129
1917	4710000	13	1944	9330000	101	1971	94000000	765	1998	144000000	2565
1918	4190000	18	1945	10900000	171	1972	101000000	914	1999	137000000	2598
1919	4150000	29	1946	15300000	178	1973	111000000	747	2000	132000000	2922
1920	6870000	17	1947	18300000	206	1974	123000000	689	2001	126000000	2860
1921	5430000	54	1948	19400000	270	1975	109000000	818	2002	136000000	3156
1922	5940000	41	1949	19700000	232	1976	109000000	908	2003	138000000	3233
1923	7120000	45	1950	23400000	228	1977	121000000	960	2004	143000000	3386
1924	7780000	35	1951	24600000	281	1978	127000000	1078	2005	150000000	3028
1925	8900000	48	1952	26400000	327	1979	134000000	929	2006	151000000	3138
1926	9380000	68	1953	27200000	311	1980	147000000	1226	2007	156000000	3520

⁷⁹ Phosphates or salts based formally on phosphorous oxoacids [128].

⁸⁰ Activity represents world production of phosphate, defined at usgs.gov as "...marketable phosphate rock. Data are from the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁸¹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Phosphate was used as the keyword found in the patent title or abstract by year of publication.

Table A3.68. Correlation Eq.(A1.1) terms calculated from Table A3.67 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
6.541E+09	85552	7.7E+17	1.63E+08	1.04E+13	3.738E+17	95356234	5.2E+12	0.870809	75.83086

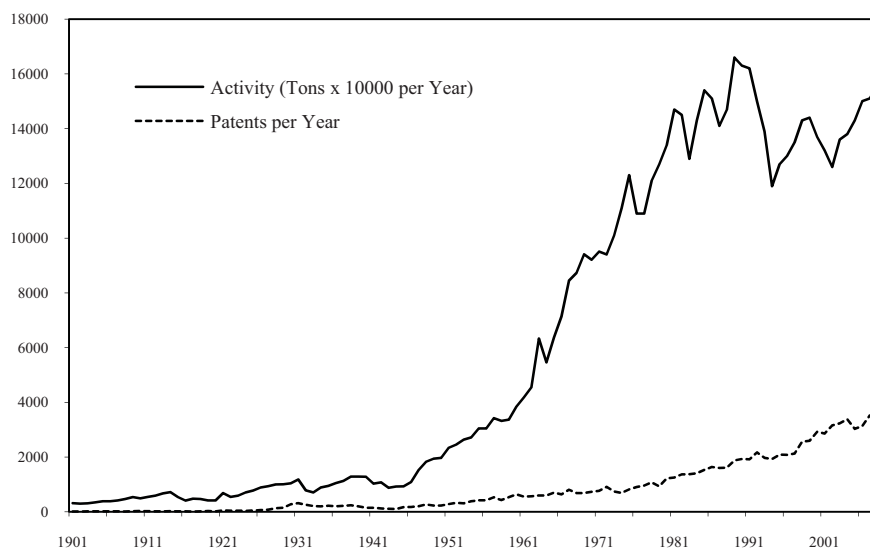


Figure A3.150. Phosphate Rock Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

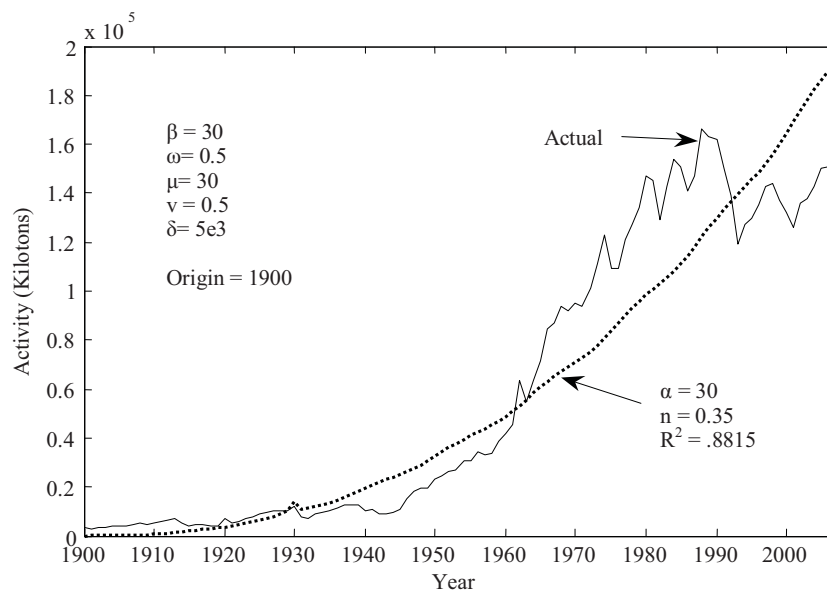


Figure A3.151. USGS World Phosphate Rock Production. World phosphate rock production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters.

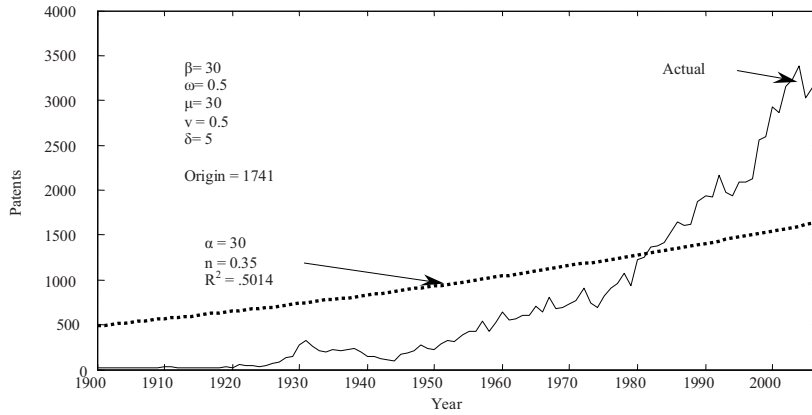


Figure A3.152. EPO Worldwide Patent Search: Phosphate in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

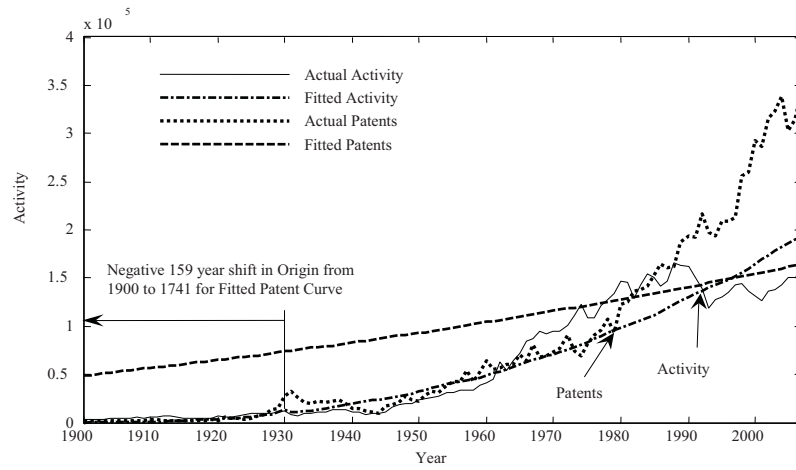


Figure A3.153. Phosphate Rock Best-fit Activity and Patents. Illustrates best-fit origin shift.

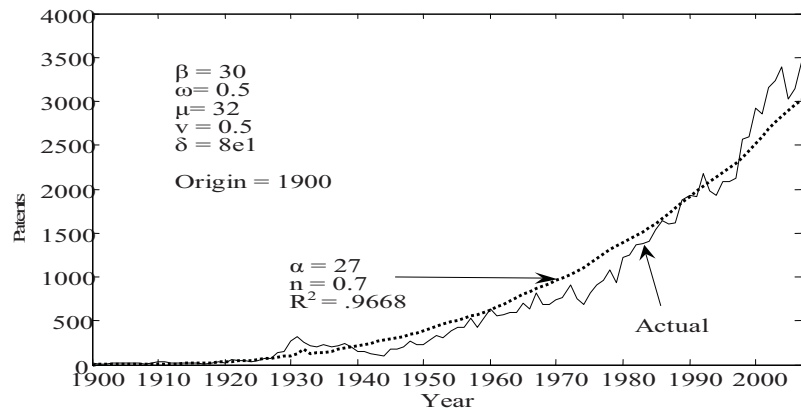


Figure A3.154. Phosphate Rock Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.69 Platinum Activity⁸² and Patents⁸³

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	6.62	57	1927	4.64	59	1954	29.2	116	1981	216	1288
1901	9.85	58	1928	4.31	71	1955	33.9	177	1982	200	1417
1902	9.33	55	1929	4.84	75	1956	34.5	204	1983	203	1525
1903	7.03	53	1930	4.75	84	1957	41.1	215	1984	238	1567
1904	9.03	54	1931	8.94	71	1958	27.7	288	1985	247	1615
1905	6.24	79	1932	6.53	66	1959	32.8	263	1986	260	2088
1906	6.59	64	1933	6.77	55	1960	39.7	364	1987	271	1811
1907	9.65	61	1934	12.9	54	1961	41.8	312	1988	280	1955
1908	8	58	1935	12.1	88	1962	50.5	293	1989	282	2466
1909	8.45	38	1936	14.2	83	1963	63.4	306	1990	291	2727
1910	8.89	46	1937	14.8	88	1964	79.2	372	1991	287	2757
1911	9.74	50	1938	16.8	100	1965	92.3	422	1992	280	3055
1912	9.77	48	1939	16.9	81	1966	94.5	405	1993	276	2687
1913	8.31	58	1940	14.5	54	1967	98.8	550	1994	269	2714
1914	8.11	44	1941	14.9	60	1968	106	515	1995	326	2646
1915	4.45	44	1942	16.9	36	1969	107	507	1996	324	2869
1916	2.8	26	1943	19.6	38	1970	132	629	1997	339	2783
1917	2.59	20	1944	16	35	1971	127	648	1998	354	3092
1918	1.96	17	1945	30	31	1972	133	758	1999	366	3078
1919	2.11	32	1946	17.9	58	1973	163	688	2000	364	3345
1920	2.3	31	1947	15.6	69	1974	179	604	2001	395	3026
1921	1.84	36	1948	16.3	105	1975	178	664	2002	414	3159
1922	2.17	53	1949	17.9	81	1976	194	740	2003	466	3011
1923	2.56	42	1950	18.7	81	1977	203	710	2004	481	2999
1924	3.56	39	1951	21	96	1978	200	805	2005	504	2828
1925	3.23	55	1952	21.8	108	1979	202	855	2006	513	2993
1926	4.42	42	1953	24.1	91	1980	213	1206	2007	509	2925

Table A3.70. Correlation Eq.(A1.1) terms calculated from Table A3.69 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
12404.68	85220	3635945	1.92E+08	25696304	2211166.3	1.25E+08	15908093	0.956927	91.57099

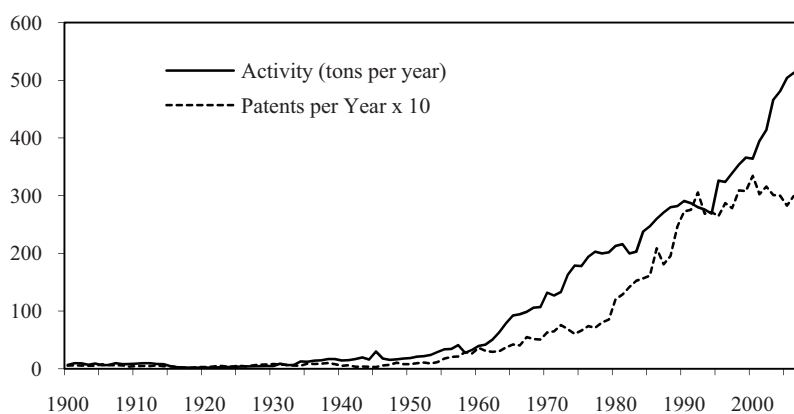


Figure A3.155. Platinum Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁸² Activity represents world production of platinum, defined at usgs.gov as “...recorded from the MR [*Mineral Resources of the United States*] and MYB [*Minerals Yearbook*] for the years 1900 to the most recent.” Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁸³ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Platinum or Pt were used as keywords found in the patent title or abstract by year of publication.

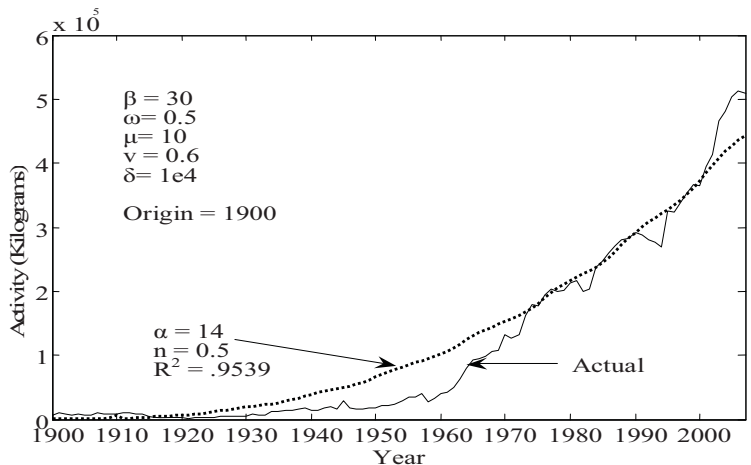


Figure A3.156. USGS World Platinum Production. World platinum production (activity) scaled to kilograms with actual and best-fit curves and common pattern equation parameters.

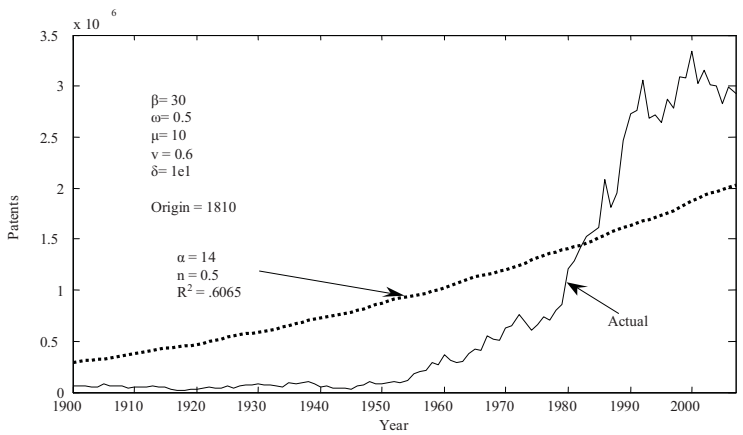


Figure A3.157. EPO Worldwide Patent Search: Platinum or Pt in Title or Abstract by date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

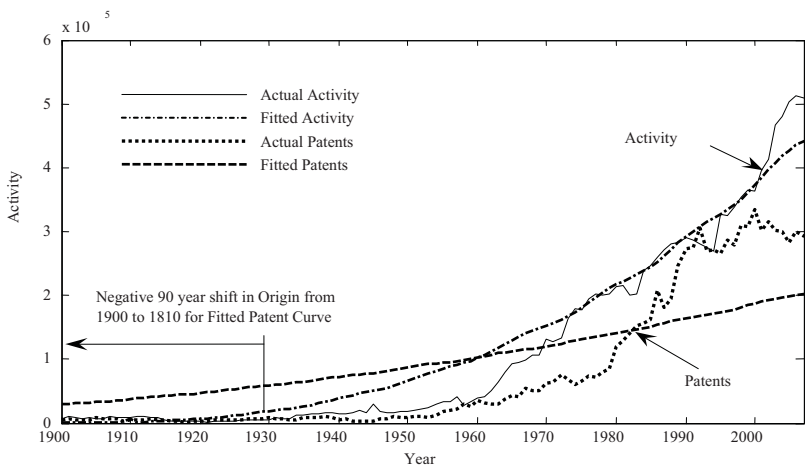


Figure A3.158. Platinum Best-fit Activity and Patents. Illustrates best-fit origin shift.

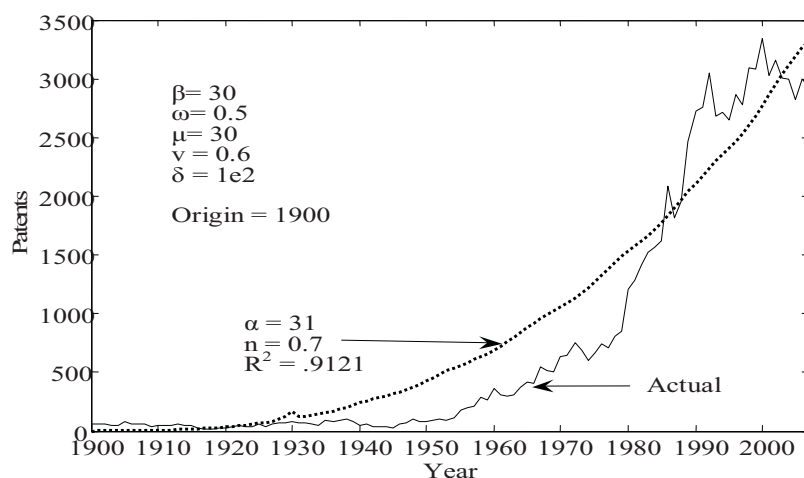


Figure A3.159. Platinum Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.71 Potash⁸⁴ Activity⁸⁵ and Patents⁸⁶

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	2000000	55	1954	6620000	35	1981	27100000	37
1901			1928	2030000	87	1955	7260000	48	1982	24500000	56
1902			1929	2200000	103	1956	7530000	56	1983	27400000	54
1903			1930	2050000	98	1957	7890000	54	1984	29300000	43
1904			1931	1400000	122	1958	7980000	40	1985	29200000	45
1905			1932	1250000	89	1959	8530000	34	1986	28800000	69
1906			1933	1670000	86	1960	9070000	33	1987	30500000	42
1907			1934	1980000	80	1961	9710000	43	1988	31800000	32
1908			1935	2270000	94	1962	9800000	42	1989	29300000	22
1909			1936	2310000	70	1963	11300000	32	1990	27500000	27
1910			1937	2820000	79	1964	12300000	28	1991	26100000	44
1911			1938	3010000	85	1965	13700000	24	1992	23900000	40
1912			1939	2730000	53	1966	14600000	22	1993	20400000	32
1913			1940	2810000	41	1967	15700000	36	1994	23100000	25
1914			1941	3210000	31	1968	16200000	33	1995	24700000	33
1915			1942	3170000	24	1969	17400000	28	1996	23900000	40
1916			1943	3270000	18	1970	18200000	30	1997	25500000	32
1917			1944	3040000	28	1971	19900000	20	1998	26000000	59
1918			1945	1910000	18	1972	20000000	25	1999	27200000	41
1919	122000	30	1946	2310000	34	1973	18900000	19	2000	27000000	61
1920	224000	32	1947	2620000	36	1974	21100000	23	2001	26400000	48
1921	994000	53	1948	2940000	50	1975	24700000	19	2002	27100000	43
1922	1400000	52	1949	2540000	26	1976	24300000	15	2003	28600000	66
1923	1250000	31	1950	3130000	23	1977	25200000	18	2004	31100000	67
1924	1100000	30	1951	5080000	31	1978	26100000	20	2005	32500000	84
1925	1590000	70	1952	5620000	49	1979	25700000	12	2006	29100000	88
1926	1710000	45	1953	5900000	46	1980	27900000	46	2007	34600000	113

⁸⁴ Potassium compounds [128].

⁸⁵ Activity represents world production of potash, defined at usgs.gov as "...recorded from the MR [*Mineral Resources of the United States*] and MYB [*Minerals Yearbook*] for the years 1900 to the most recent." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁸⁶ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Potash was used as the keyword found in the patent title or abstract by year of publication.

Table A3.72. Correlation Eq.(A1.1) terms calculated from Table A3.71 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
1.254E+09	4077	2.89E+16	237793	5.4E+10	1.127E+16	51029.75	-3.4E+09	-0.14144	2.000428

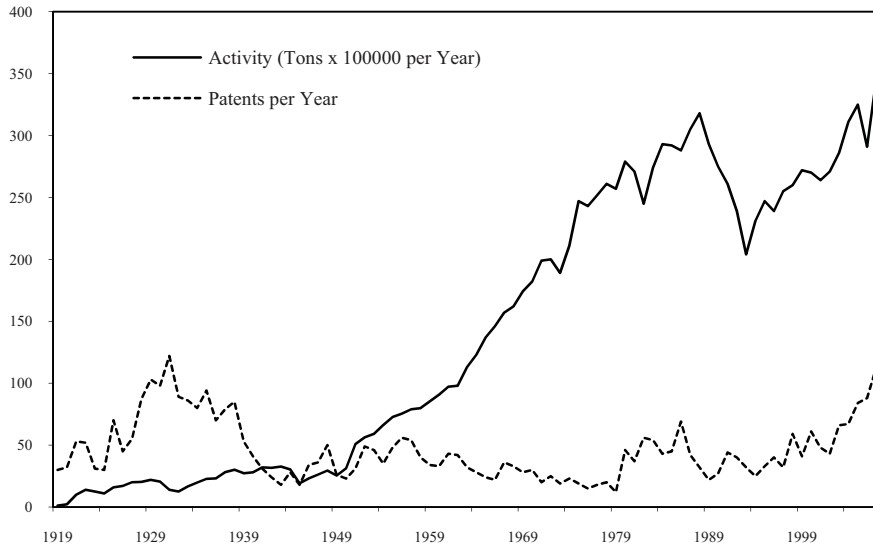


Figure A3.160. Potash Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

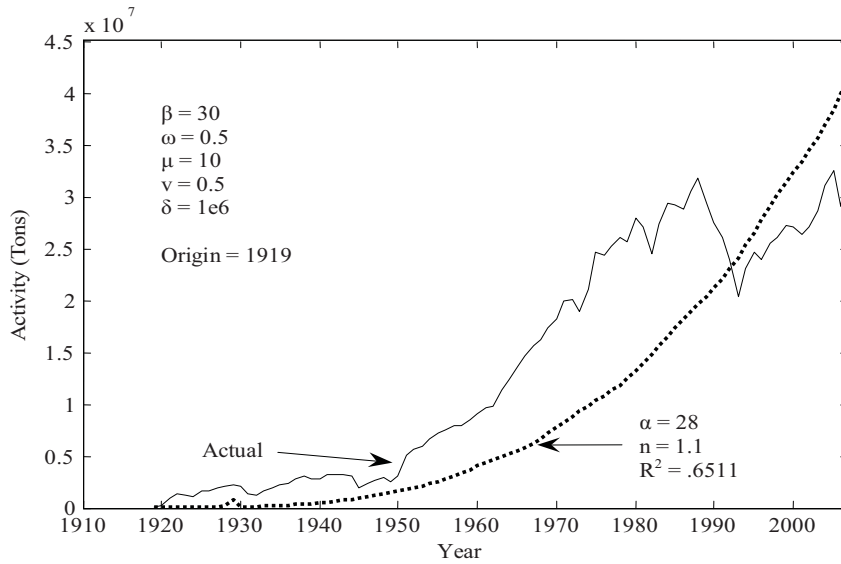


Figure A3.161. USGS World Potash production. World potash production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

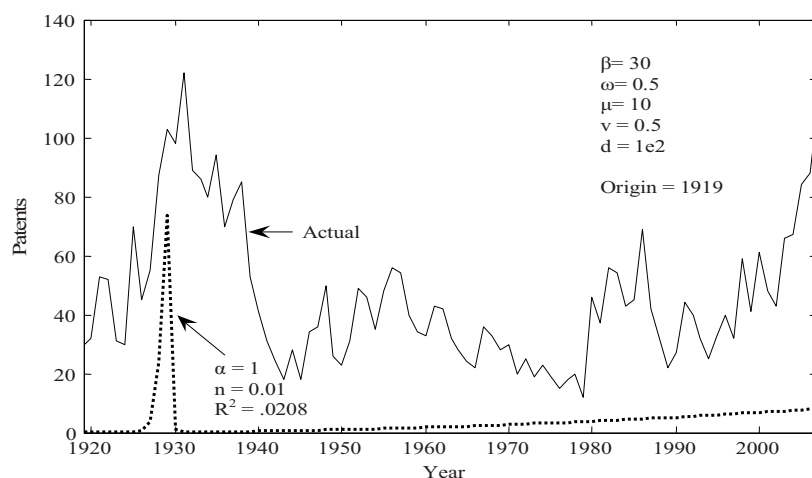


Figure A3.162. Potash Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.73 Rare Earths⁸⁷ Activity⁸⁸ and Patents⁸⁹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	1040	3	1927	352	2	1954	7840	3	1981	30600	229
1901	1090	1	1928	180	4	1955	5760	7	1982	26600	229
1902	863	2	1929	197	3	1956	5230	6	1983	31400	204
1903	2030	1	1930	17	8	1957	5980	16	1984	41400	212
1904	2860	4	1931	50	1	1958	8060	8	1985	43500	223
1905	2780	0	1932	530	3	1959	2810	9	1986	39900	267
1906	2600	2	1933	302	8	1960	2270	30	1987	46900	328
1907	2580	3	1934	564	1	1961	3690	23	1988	55300	429
1908	2840	2	1935	2130	4	1962	8020	24	1989	60700	762
1909	3690	0	1936	1840	4	1963	6060	28	1990	52900	551
1910	3020	0	1937	2150	6	1964	3680	27	1991	41700	496
1911	2490	0	1938	3310	4	1965	6960	35	1992	50100	506
1912	2500	1	1939	2510	2	1966	16200	44	1993	46700	554
1913	1480	0	1940	2370	2	1967	16900	56	1994	55100	497
1914	992	2	1941	2380	2	1968	16200	78	1995	74300	503
1915	870	1	1942	1500	1	1969	18100	75	1996	79700	482
1916	731	0	1943	1900	0	1970	15900	89	1997	68300	509
1917	1730	0	1944	3200	2	1971	16400	83	1998	77100	566
1918	1470	0	1945	1440	2	1972	18200	92	1999	86600	605
1919	1210	0	1946	721	0	1973	24000	106	2000	90900	714
1920	1590	0	1947	1300	7	1974	25600	82	2001	94500	742
1921	929	1	1948	2720	5	1975	22100	82	2002	98200	811
1922	189	3	1949	1290	1	1976	19700	100	2003	97100	853
1923	138	5	1950	470	4	1977	24500	113	2004	102000	851
1924	348	1	1951	1240	3	1978	26500	163	2005	122000	792
1925	12	2	1952	1820	2	1979	28800	131	2006	137000	822
1926	146	2	1953	3960	3	1980	27300	196	2007	124000	983

⁸⁷ Rare earth oxides [78]. Lanthanide, lanthanoid, yttrium or scandium.

⁸⁸ Activity represents world production of the rare earths, defined at usgs.gov as "...REO (Rare Earth Oxides) content of ores produced." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁸⁹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Lanthanide, lanthanoid, yttrium or scandium were used as keywords found in the patent title or abstract by year of publication.

Table A3.74. Correlation Eq.(A1.1) terms calculated from Table A3.73 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
2339921	16546	1.62E+11	9589606	1.22E+09	1.116E+11	7054698	8.61E+08	0.96997	94.08413

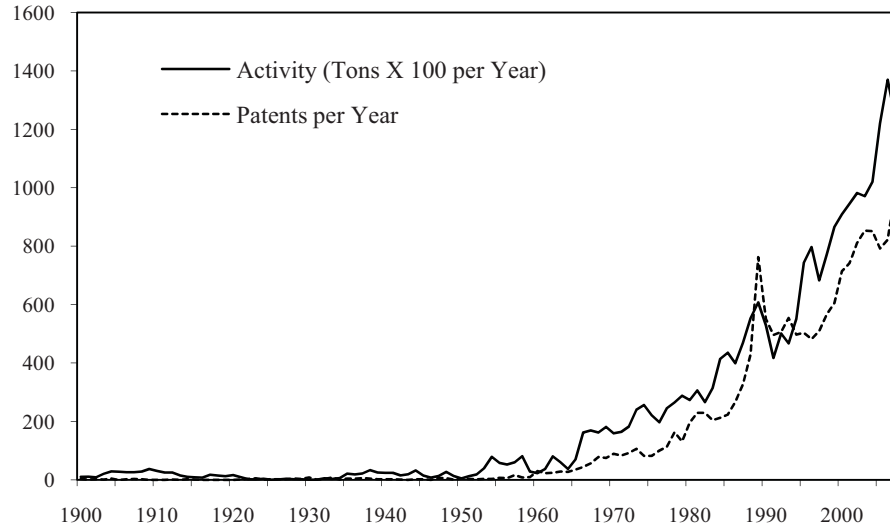


Figure A3.163. Rare Earths Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

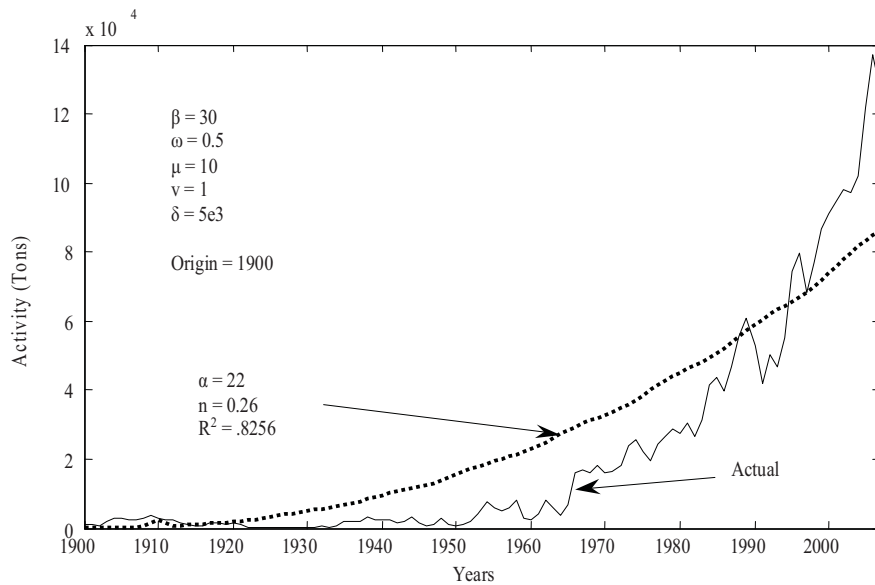


Figure A3.164. USGS World Rare Earth Production. World rare earths production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

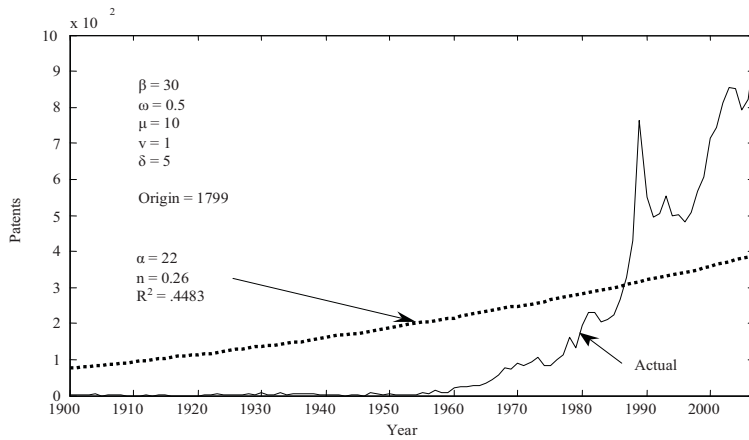


Figure A3.165. EPO Worldwide Patent Search: Lanthanide, Lanthanoid, Yttrium or Scandium in Title or Abstract by date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

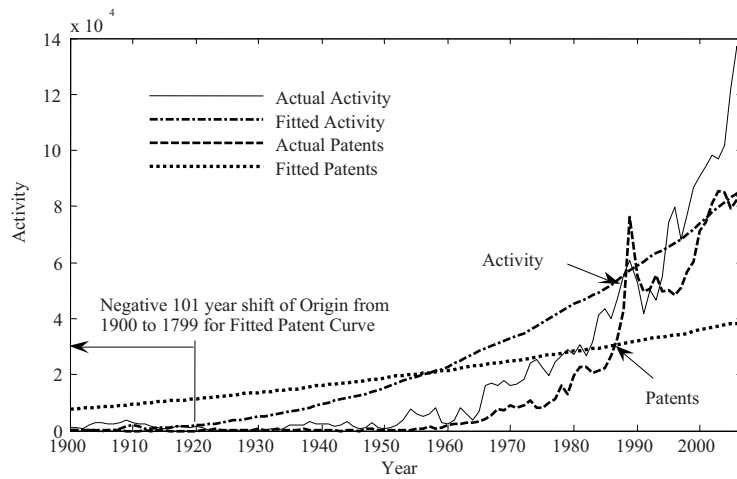


Figure A3.166. Rare Earths Best-Fit Activity and Patents. Illustrates best-fit origin shift.

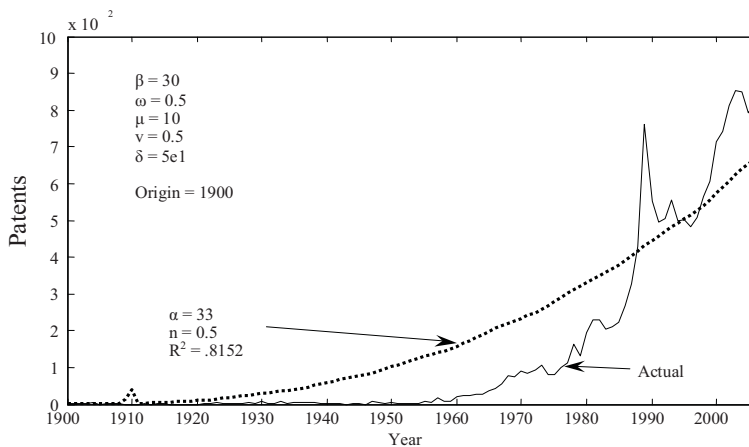


Figure A3.167. Rare Earths Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.75 Salt Activity⁹⁰ and Patents⁹¹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (pat)	Yr.	x (activity)	y (patent)
1900			1927	26100000	291	1954	60500000	836	1981	171000000	6311
1901			1928	27100000	343	1955	65000000	997	1982	164000000	6632
1902			1929	25700000	431	1956	68200000	1169	1983	159000000	6703
1903			1930	28100000	463	1957	71900000	1291	1984	173000000	7003
1904			1931	24400000	560	1958	74800000	1100	1985	173000000	7234
1905			1932	26200000	485	1959	79700000	1175	1986	175000000	7875
1906			1933	28500000	440	1960	84800000	1606	1987	179000000	7890
1907			1934	30200000	492	1961	85000000	1411	1988	186000000	8081
1908			1935	30700000	559	1962	91500000	1374	1989	192000000	9163
1909			1936	31800000	579	1963	96100000	1514	1990	183000000	9630
1910			1937	30200000	579	1964	98600000	1692	1991	202000000	9151
1911			1938	27900000	703	1965	109000000	2088	1992	185000000	9830
1912			1939	32000000	498	1966	111000000	1782	1993	187000000	9829
1913	17600000	153	1940	33200000	435	1967	119000000	2420	1994	191000000	10815
1914	16900000	140	1941	36700000	335	1968	126000000	2186	1995	199000000	11042
1915	15300000	116	1942	38600000	272	1969	137000000	1960	1996	204000000	10836
1916	17100000	73	1943	41200000	261	1970	146000000	2442	1997	221000000	11252
1917	17600000	65	1944	40400000	267	1971	144000000	2309	1998	200000000	12879
1918	17800000	90	1945	36000000	310	1972	146000000	3041	1999	210000000	12277
1919	19800000	123	1946	38300000	350	1973	155000000	2607	2000	195000000	14103
1920	21900000	127	1947	40500000	455	1974	166000000	2368	2001	199000000	13958
1921	17600000	202	1948	44000000	665	1975	162000000	2996	2002	214000000	15031
1922	23400000	186	1949	43000000	571	1976	161000000	3609	2003	225000000	15927
1923	23300000	168	1950	48100000	548	1977	157000000	3707	2004	236000000	16829
1924	23700000	208	1951	55900000	664	1978	168000000	3929	2005	250000000	16158
1925	25000000	260	1952	54200000	812	1979	173000000	4069	2006	262000000	15119
1926	26200000	247	1953	59300000	689	1980	169000000	5833	2007	257000000	17405

Table A3.76. Correlation Eq.(A1.1) terms calculated from Table A3.76 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
9.879E+09	375689	1.56E+18	3.8E+09	7.08E+13	5.359E+17	2.32E+09	3.17E+13	0.899574	80.92342

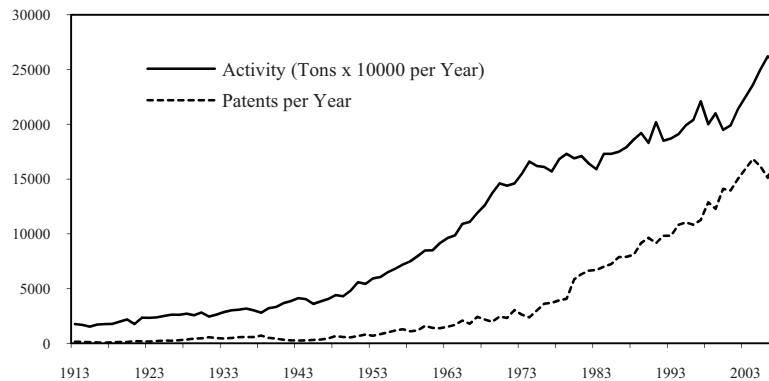


Figure A3.168. Salt Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁹⁰ Activity represents world production of salt, defined at usgs.gov as "...reported in the MR for the years 1900–06. The MR stopped reporting world production for the years 1907–22. The 1923 MR reported world salt production for the years 1913–23 with continued reporting in the MR [*Minerals resources of the United states*] and in the MYB [*Minerals Yearbook*] for the years 1924 to the most recent." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁹¹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Salt was used as the keyword found in the patent title or abstract by year of publication.

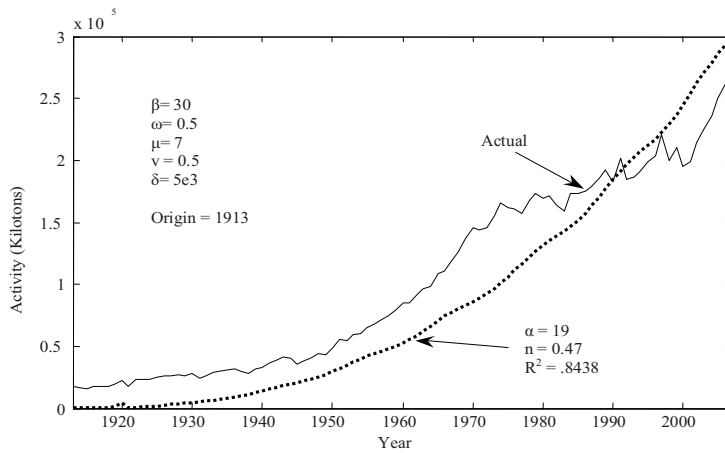


Figure A3.169. USGS World Salt Production. World salt production (activity) scaled to metric kilotons with actual and best-fit curves and common pattern equation parameters.

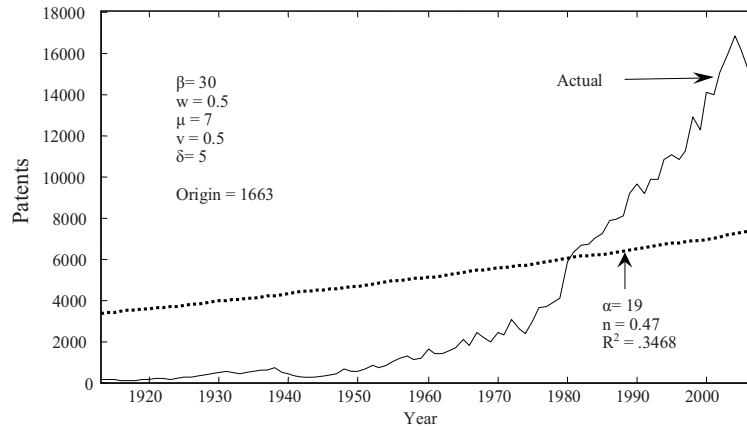


Figure A3.170. EPO Worldwide Patent Search: Salt in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

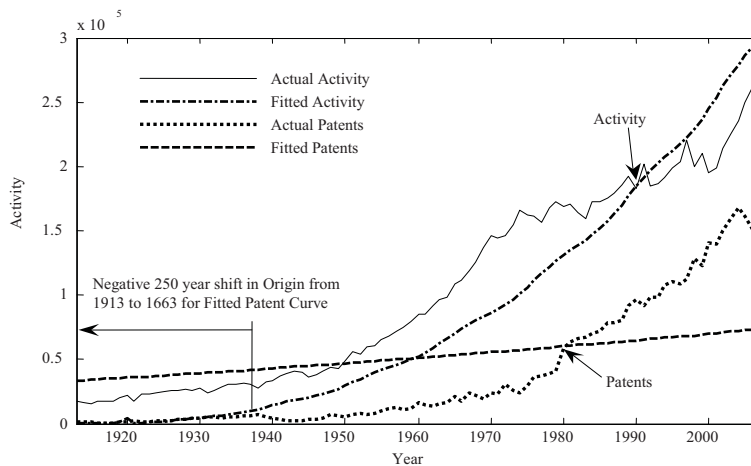


Figure A3.171. Salt Best-Fit Activity and Patents. Illustrates best-fit origin shift.

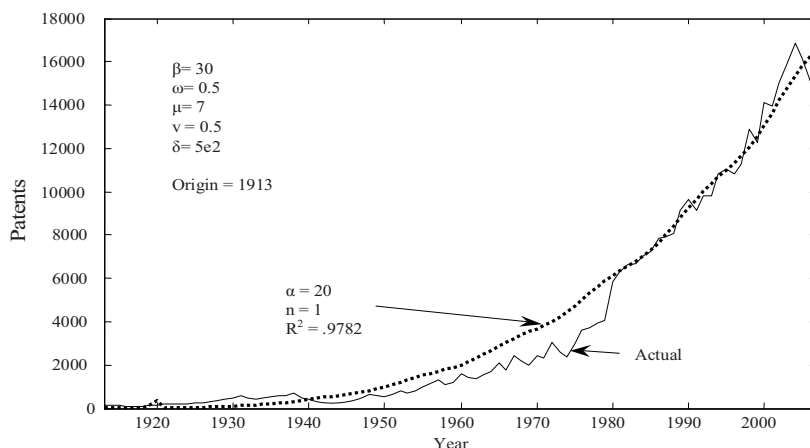


Figure A3.172. Salt Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.77 Selenium Activity⁹² and Patents⁹³

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	636	77	1981	1290	173
1901			1928			1955	736	72	1982	1120	212
1902			1929			1956	872	90	1983	1400	206
1903			1930			1957	872	86	1984	1490	168
1904			1931			1958	663	92	1985	1320	199
1905			1932			1959	748	83	1986	1400	184
1906			1933			1960	758	102	1987	1420	201
1907			1934			1961	951	92	1988	1680	230
1908			1935			1962	948	107	1989	1600	261
1909			1936			1963	914	97	1990	1770	206
1910			1937			1964	981	113	1991	1640	193
1911			1938	285	75	1965	816	96	1992	1770	253
1912			1939	194	40	1966	895	80	1993	1740	220
1913			1940	251	43	1967	930	104	1994	2160	249
1914			1941	754	38	1968	883	82	1995	2070	260
1915			1942	645	42	1969	1290	77	1996	2250	267
1916			1943	532	45	1970	1310	94	1997	1720	317
1917			1944	424	55	1971	1140	84	1998	1470	310
1918			1945	387	48	1972	1230	102	1999	1410	323
1919			1946	475	52	1973	1220	103	2000	1460	304
1920			1947	508	77	1974	1210	89	2001	1470	305
1921			1948	471	117	1975	1180	106	2002	1480	353
1922			1949	387	97	1976	1110	117	2003	1570	400
1923			1950	418	66	1977	1380	148	2004	1440	453
1924			1951	488	55	1978	1440	157	2005	1340	500
1925			1952	532	66	1979	1620	132	2006	1440	385
1926			1953	668	60	1980	1280	184	2007	1470	534

⁹² Activity represents world production of selenium, defined at usgs.gov as "...world refinery production of selenium metal. Data were not available for the years 1900–37. World production estimates for the years 1985–1987 and 1997 to the most recent do not include withheld U.S. production data. Data were recorded from the MR [*Minerals Resources of the United States*] and MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁹³ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Selenium was used as the keyword found in the patent title or abstract by year of publication.

Table A3.78. Correlation Eq.(A1.1) terms calculated from Table A3.77 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
77822	11408	1.03E+08	2800784	15380861	16376463	941605.9	2698098	0.687089	47.20916

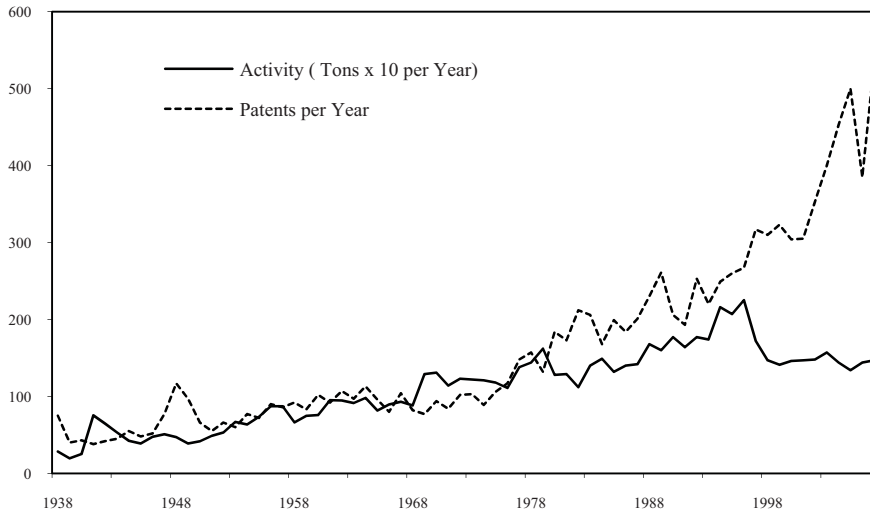


Figure A3.173. Selenium Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

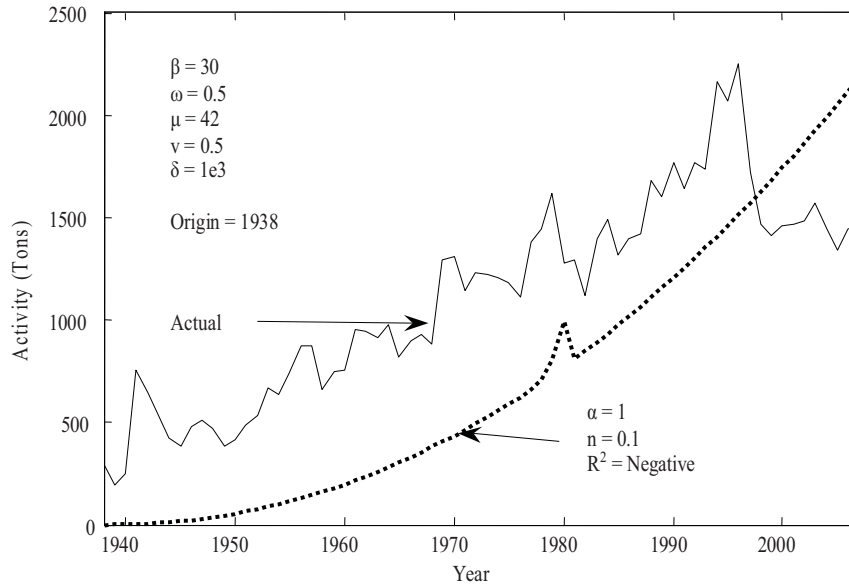


Figure A3.174. USGS World Selenium Production. World selenium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. The negative R^2 may indicate a possible Stage IV. No best-fit for the patent data was obtainable.

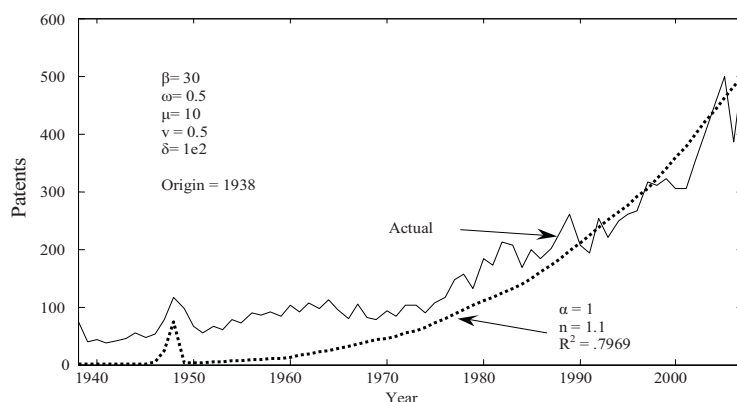


Figure A3.175. Selenium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.79 Silicon Activity⁹⁴ and Patents⁹⁵

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981	2600000	7028
1901			1928			1955			1982	2410000	8852
1902			1929			1956			1983	2540000	9680
1903			1930			1957			1984	2730000	10189
1904			1931			1958			1985	2830000	11640
1905			1932			1959			1986	2740000	13130
1906			1933			1960			1987	2760000	13268
1907			1934			1961			1988	2990000	13966
1908			1935			1962			1989	3380000	14889
1909			1936			1963			1990	4130000	14900
1910			1937			1964	1130000	995	1991	3950000	14050
1911			1938			1965	1160000	1225	1992	3470000	16106
1912			1939			1966	1160000	1157	1993	3200000	15057
1913			1940			1967	1490000	1387	1994	3170000	15032
1914			1941			1968	1540000	1345	1995	3100000	14428
1915			1942			1969	1590000	1277	1996	3200000	14643
1916			1943			1970	1640000	1485	1997	3400000	14422
1917			1944			1971	1570000	1510	1998	3200000	16633
1918			1945			1972	1670000	2058	1999	3400000	17869
1919			1946			1973	1780000	1752	2000	3500000	22557
1920			1947			1974	1800000	1516	2001	3500000	23289
1921			1948			1975	2100000	1730	2002	3720000	24647
1922			1949			1976	2320000	2029	2003	4500000	22817
1923			1950			1977	2260000	2698	2004	5030000	22574
1924			1951			1978	2550000	3071	2005	5160000	21012
1925			1952			1979	2840000	4126	2006	5400000	21762
1926			1953			1980	2750000	6338	2007	5590000	22494

⁹⁴ Activity represents world production of silicon, defined at usgs.gov as "...for the years 1964–78 were reported in the MFP [*Mineral Facts and Problems*]. World production data for the years 1979 to the most recent were reported in the MCS [*Mineral Commodity Summaries*]. World production data for the years 1964–2005 represent the total silicon content in all ferrosilicon and 3 silicon metal that were produced annually, excluding silicon metal production in China. Starting in 2006, world production data exclude the amount of silicon metal that was produced annually in the United States. Global silicon metal production data were found on a gross-weight basis in the ferroalloys chapter of the MYB [*Minerals Yearbook*]; the typical silicon content of silicon metal is 98% of the gross weight." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁹⁵ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Silicon or Si were used as keywords found in the patent title or abstract by year of publication.

Table A3.80. Correlation Eq.(A1.1) terms calculated from Table A3.79 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
126950000	472633	4.22E+14	7.8E+09	1.71E+12	5.528E+13	2.72E+09	3.45E+11	0.889319	79.08883

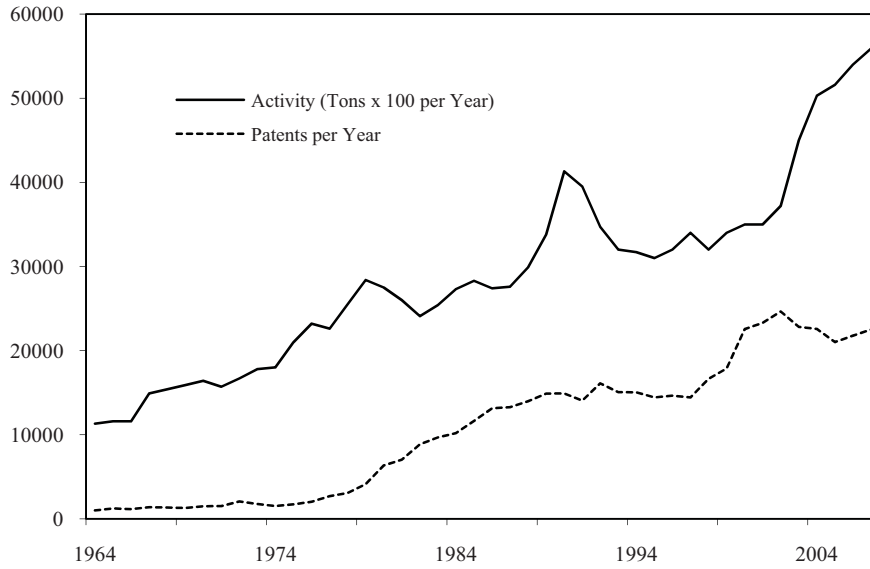


Figure A3.176. Silicon Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

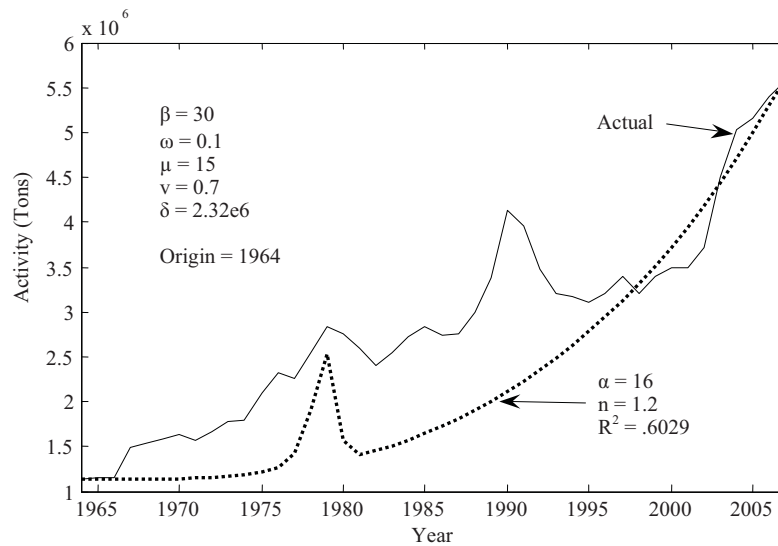


Figure A3.177. USGS World Silicon production. World silicon production scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

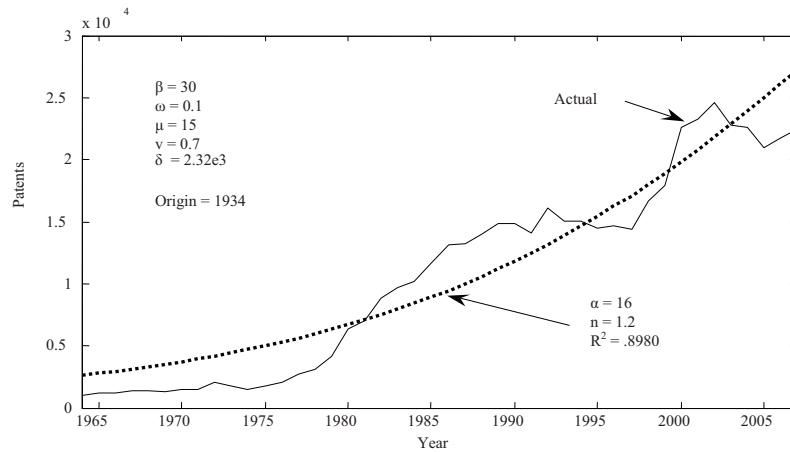


Figure A3.178. EPO Worldwide Patent Search: Silicon or Si in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

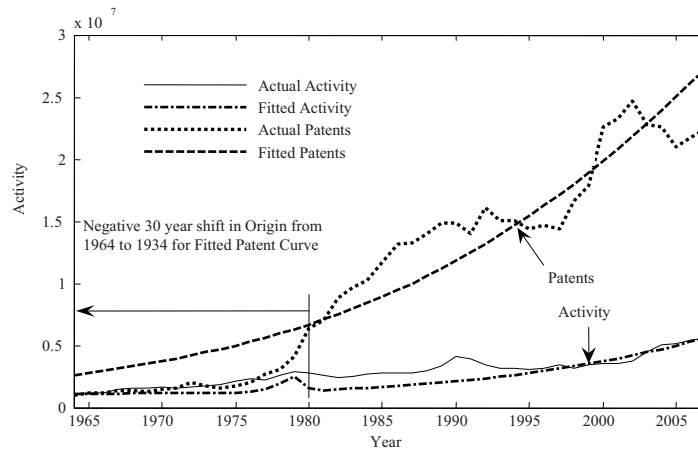


Figure A3.179. Silicon Best-Fit Activity and Patents. Illustrates best-fit origin shift.

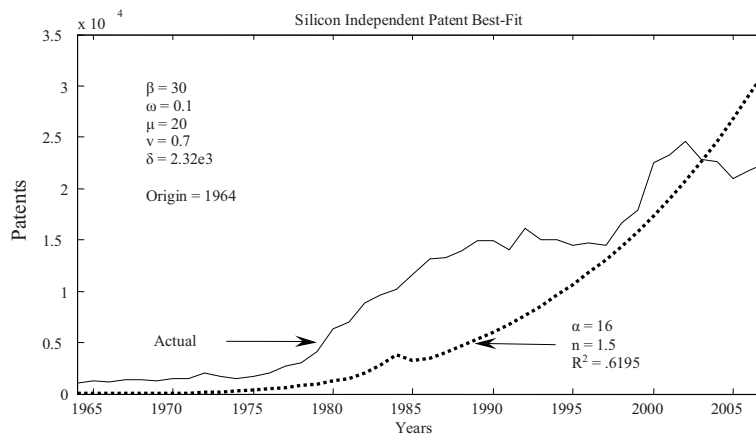


Figure A3.180. Silicon Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.81 Silver Activity⁹⁶ and Patents⁹⁷

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	5400	62	1927	7900	146	1954	6670	422	1981	11200	2797
1901	5380	58	1928	8020	184	1955	7000	464	1982	11500	2975
1902	5060	57	1929	8120	227	1956	7020	475	1983	12100	3264
1903	5220	77	1930	7740	208	1957	7190	555	1984	13100	3686
1904	5110	59	1931	6080	230	1958	7430	577	1985	13100	3824
1905	5360	74	1932	5130	220	1959	6910	654	1986	13000	4516
1906	5130	53	1933	5340	216	1960	7320	718	1987	14000	5050
1907	5730	75	1934	5990	260	1961	7370	818	1988	15500	5709
1908	6320	70	1935	6890	296	1962	7650	867	1989	16400	5736
1909	6600	77	1936	7920	371	1963	7780	898	1990	16600	6312
1910	6900	66	1937	8640	373	1964	7730	1023	1991	15600	5726
1911	7040	63	1938	8320	261	1965	8010	1140	1992	14900	5940
1912	6980	71	1939	8300	211	1966	8300	1207	1993	14100	5999
1913	7010	69	1940	8570	212	1967	8030	1209	1994	14000	5831
1914	5240	57	1941	8140	212	1968	8560	1954	1995	14900	5995
1915	5730	43	1942	7780	262	1969	9200	4102	1996	15100	5994
1916	5250	39	1943	6380	276	1970	9360	5156	1997	16500	6264
1917	5420	49	1944	7540	310	1971	9170	4980	1998	17200	5775
1918	6140	44	1945	5040	237	1972	9380	5256	1999	17600	5918
1919	5490	83	1946	3970	318	1973	9700	5373	2000	18100	6139
1920	5390	72	1947	5220	254	1974	9260	4742	2001	18900	6346
1921	5330	73	1948	5440	251	1975	9430	3269	2002	18800	6040
1922	6530	84	1949	5570	295	1976	9840	2089	2003	18800	6119
1923	7650	86	1950	6320	292	1977	10300	1434	2004	19900	5528
1924	7450	93	1951	6210	357	1978	10700	1953	2005	20600	5249
1925	7650	119	1952	6700	369	1979	10800	2322	2006	20200	4230
1926	7890	137	1953	6900	427	1980	10700	2356	2007	20800	2860

Table A3.82 Correlation Eq.(A1.1) terms calculated from Table A3.81 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
1011880	205020	1.15E+10	9.55E+08	2.85E+09	1.998E+09	5.66E+08	9.29E+08	0.873549	76.30875

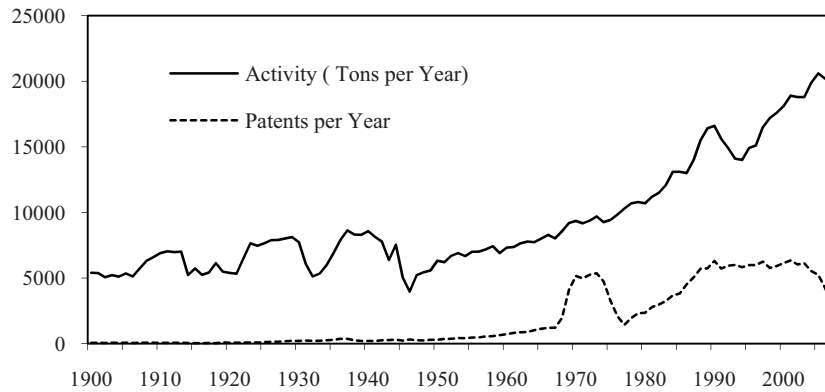


Figure A3.181. Silver Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

⁹⁶ Activity represents world production of silver, defined at usgs.gov as "...for the years 1900 to the most recent represent the recoverable silver content of precious-metal ores that were extracted from mines throughout the world. World production data were recorded from the MR [Minerals Resources of the United States] and MYB [Minerals Yearbook]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁹⁷ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Silver or Ag were used as keywords found in the patent title or abstract by year of publication.

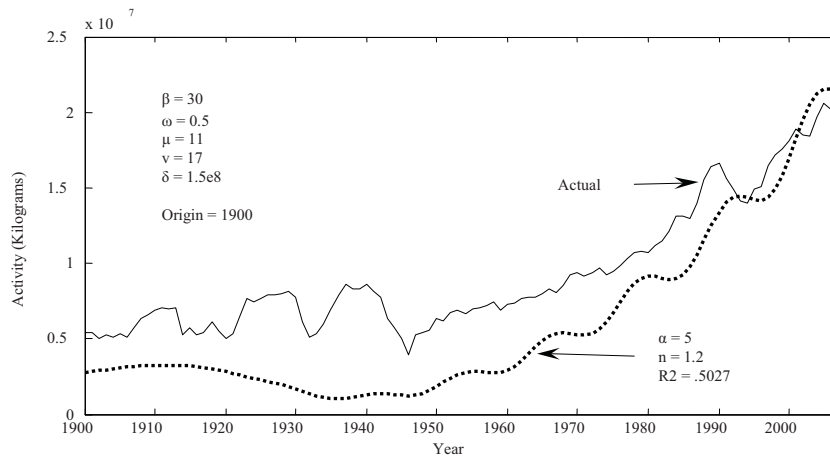


Figure A3.182. USGS World Silver Production. World silver production (activity) scaled in kilograms with actual and best-fit curves and common pattern equation parameters.

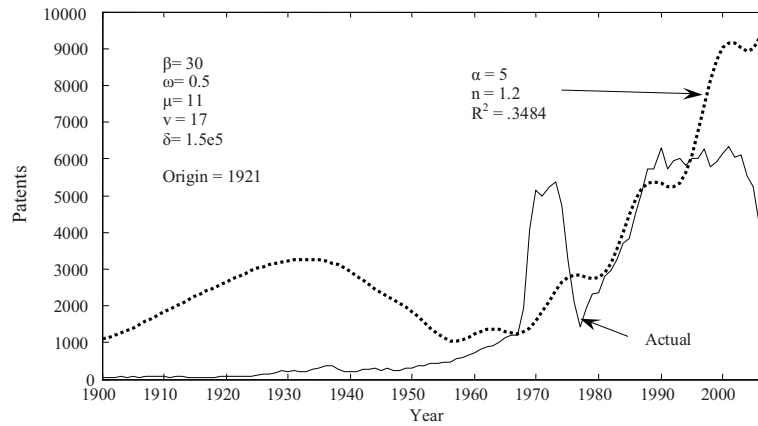


Figure A3.183. EPO Worldwide Patent Search: Silver or Ag in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

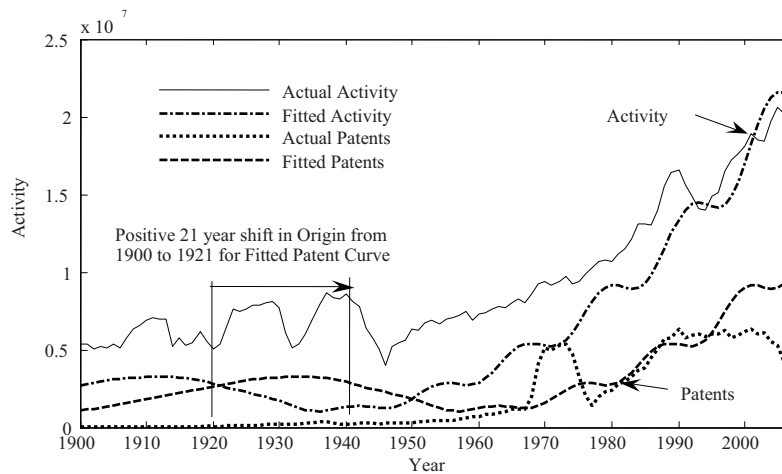


Figure A3.184. Silver Best-Fit Activity and Patents. Illustrates best-fit origin shift.

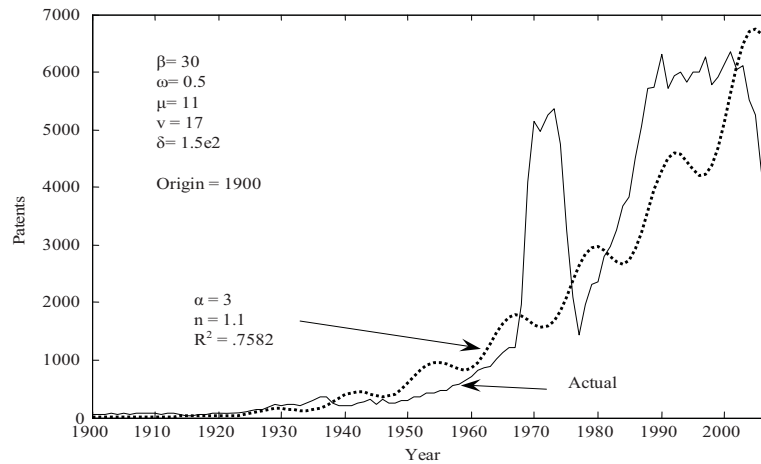


Figure A3.185. Silver Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.83 Sulfur Activity⁹⁸ and Patents⁹⁹

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	1420000	101	1927	5890000	214	1954	12800000	474	1981	53600000	2229
1901	1420000	80	1928	5690000	285	1955	15500000	462	1982	50600000	2125
1902	4390000	117	1929	6200000	334	1956	17000000	606	1983	49800000	2010
1903	4540000	114	1930	6400000	392	1957	17000000	613	1984	52500000	1970
1904	1320000	70	1931	5180000	499	1958	16100000	532	1985	53800000	2008
1905	1420000	79	1932	3760000	408	1959	17200000	539	1986	53700000	2206
1906	1320000	89	1933	4570000	409	1960	19600000	675	1987	57000000	2284
1907	1320000	83	1934	5080000	364	1961	20600000	579	1988	59200000	2457
1908	1220000	95	1935	5690000	392	1962	21300000	643	1989	58900000	2620
1909	1320000	84	1936	5390000	385	1963	21900000	689	1990	57800000	2654
1910	1420000	110	1937	5990000	412	1964	23500000	745	1991	54600000	2789
1911	1420000	101	1938	5590000	440	1965	25200000	805	1992	50700000	3080
1912	1630000	106	1939	7320000	316	1966	26500000	741	1993	51600000	2896
1913	1830000	97	1940	7930000	237	1967	28400000	815	1994	53400000	3082
1914	1070000	85	1941	7520000	201	1968	29500000	857	1995	54800000	2919
1915	1290000	67	1942	8030000	182	1969	30700000	743	1996	55200000	3000
1916	1440000	37	1943	6710000	122	1970	41900000	851	1997	56900000	3299
1917	1980000	56	1944	6600000	148	1971	42700000	854	1998	57400000	3563
1918	2140000	56	1945	6200000	209	1972	45500000	1081	1999	57400000	3652
1919	1800000	89	1946	7320000	260	1973	48200000	1075	2000	59300000	4182
1920	1590000	108	1947	8640000	302	1974	51200000	1081	2001	59500000	4079
1921	2230000	172	1948	9450000	331	1975	50700000	1470	2002	62000000	4535
1922	2090000	160	1949	9960000	328	1976	50900000	1745	2003	64100000	4410
1923	2380000	151	1950	10800000	299	1977	52300000	1698	2004	66200000	4523
1924	3860000	176	1951	11400000	347	1978	52100000	1776	2005	67000000	4288
1925	4780000	190	1952	12100000	433	1979	53200000	1687	2006	66800000	3766
1926	5490000	195	1953	11700000	368	1980	55000000	2192	2007	68400000	4181

⁹⁸ Activity represents world production of sulfur, defined at usgs.gov as "...all forms of sulfur and are in terms of their sulfur content. Data prior to 1936 include elemental sulfur production from principal producing countries and world pyrite production. Data for the years 1936 to the most recent are world production of all forms of sulfur. Data are from the MR [*Minerals Resources of the United States*] and MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

⁹⁹ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Sulfur or Sulphur were used as keywords found in the patent title or abstract by year of publication.

Table A3.84 Correlation Eq.(A1.1) terms calculated from Table A3.83 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
2.694E+09	122320	1.27E+17	3.2E+08	6.06E+12	5.937E+16	1.81E+08	3.01E+12	0.915594	83.83128

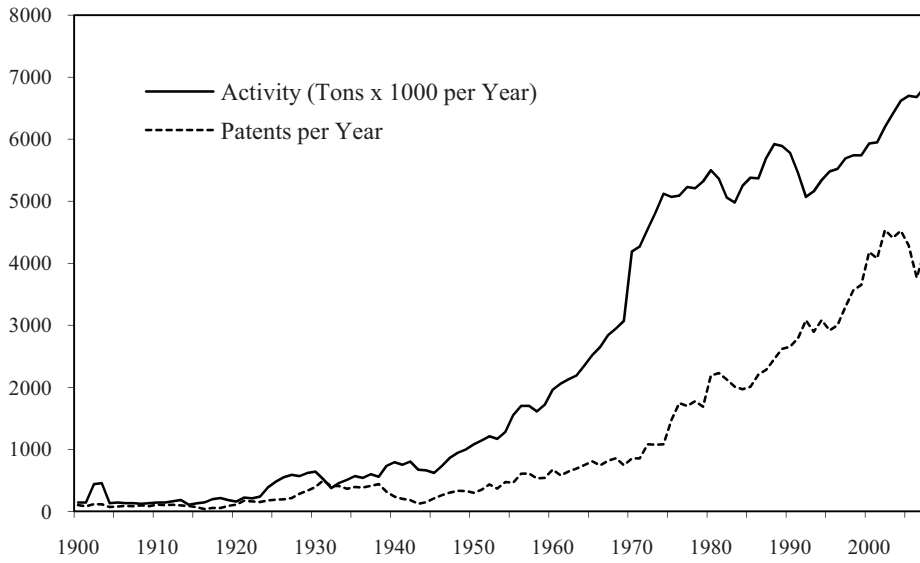


Figure A3.186. Sulfur Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

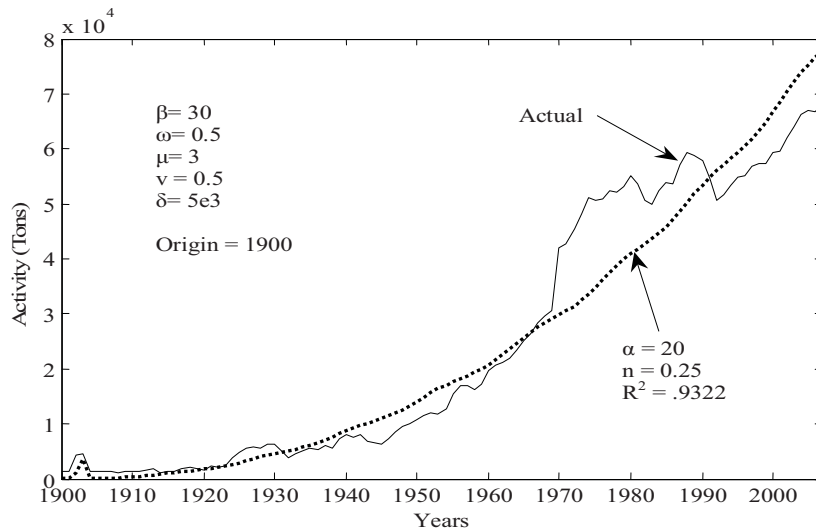


Figure A3.187. USGS World Sulfur Production. World sulfur production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters.

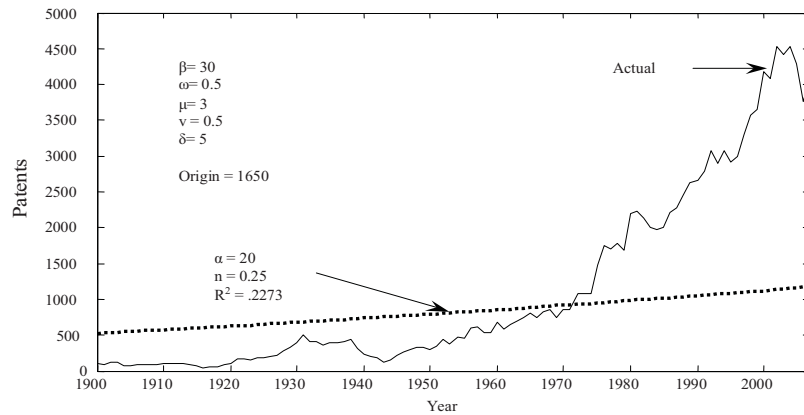


Figure A3.188. EPO Worldwide Patent Search: Sulfur or Sulphur in Title or Abstract by date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

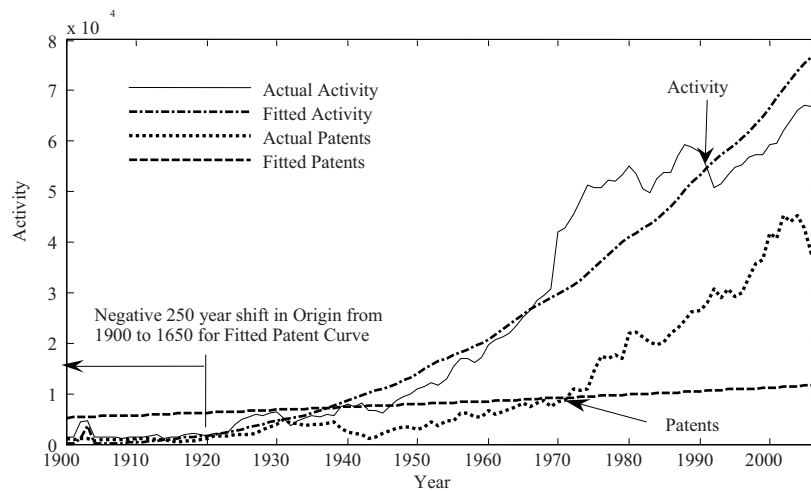


Figure A3.189. Sulfur Best-Fit Activity and Patents. Illustrates best-fit origin shift.

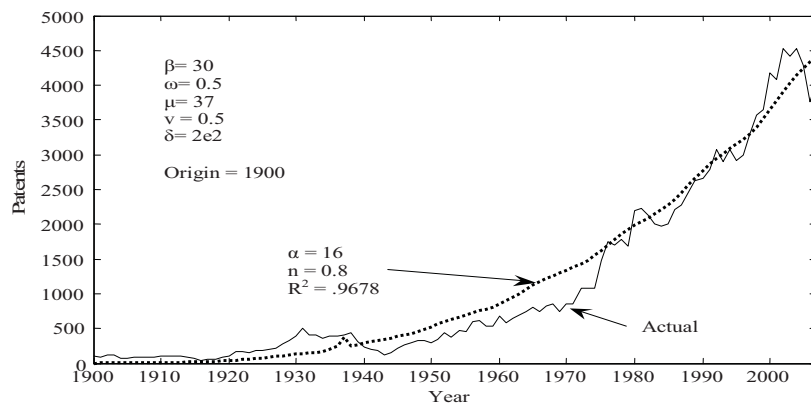


Figure A3.190. Sulfur Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.85 Talc and Pyrophyllite¹⁰⁰ Activity¹⁰¹ and Patents¹⁰²

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	431000	18	1954	1470000	59	1981	7270000	198
1901			1928	389000	26	1955	1620000	88	1982	7060000	212
1902			1929	421000	24	1956	1750000	98	1983	7060000	249
1903			1930	364000	30	1957	2010000	131	1984	7570000	246
1904	118000	11	1931	384000	40	1958	1910000	98	1985	7830000	271
1905	124000	9	1932	331000	32	1959	2350000	159	1986	7760000	285
1906	151000	12	1933	430000	35	1960	2520000	153	1987	8470000	264
1907	191000	13	1934	399000	36	1961	2710000	129	1988	8810000	285
1908	160000	16	1935	424000	51	1962	2670000	110	1989	9240000	361
1909	178000	13	1936	472000	53	1963	2990000	160	1990	9370000	315
1910	202000	9	1937	515000	45	1964	3520000	143	1991	9060000	317
1911	208000	11	1938	420000	49	1965	3570000	167	1992	8500000	349
1912	171000	15	1939	488000	38	1966	3710000	131	1993	8420000	340
1913	279000	10	1940	664000	33	1967	3960000	162	1994	8260000	323
1914	213000	11	1941	840000	26	1968	4350000	119	1995	8490000	365
1915	224000	3	1942	1170000	20	1969	4680000	107	1996	9880000	371
1916	257000	5	1943	1120000	13	1970	4820000	128	1997	10400000	379
1917	266000	4	1944	1010000	12	1971	4740000	138	1998	9410000	445
1918	252000	4	1945	840000	20	1972	4830000	169	1999	9470000	424
1919	255000	7	1946	950000	28	1973	5400000	114	2000	8730000	497
1920	322000	10	1947	1060000	33	1974	5810000	105	2001	9060000	401
1921	207000	12	1948	1300000	46	1975	4900000	120	2002	8030000	446
1922	353000	16	1949	1280000	39	1976	5270000	143	2003	7800000	389
1923	336000	14	1950	1430000	41	1977	6090000	118	2004	7840000	393
1924	375000	8	1951	1570000	42	1978	6400000	141	2005	7950000	289
1925	398000	10	1952	1410000	54	1979	6870000	125	2006	7750000	263
1926	344000	17	1953	1480000	51	1980	7540000	208	2007	7620000	338

Table A3.86 Correlation Eq.(A1.1) terms calculated from Table A3.85 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
362746000	13713	2.47E+15	3678389	9.23E+10	1.2E+15	1870251	4.45E+10	0.938294	88.03948

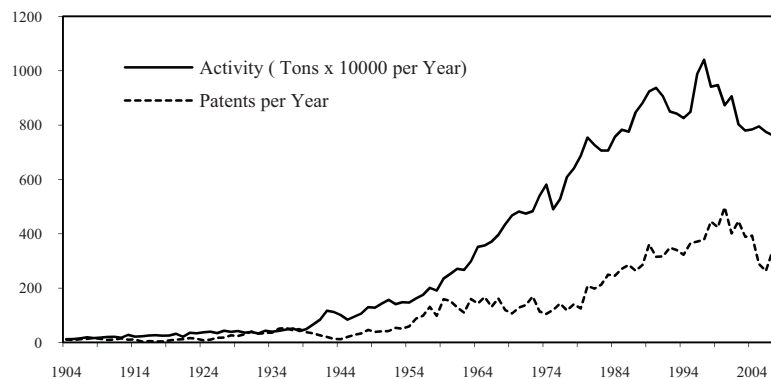


Figure A3.191. Talc and Pyrophyllite Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

¹⁰⁰ Magnesium silicate and aluminum silicate hydroxide [78,128].

¹⁰¹ Activity represents world production of talc, defined at usgs.gov as "...for the years 1904 to the most recent were recorded from the MR [Minerals Resources of the United States] and MYB [Minerals Yearbook]. World production data for the years 1904–12 represent the summed weights of all talc and soapstone materials that were produced annually throughout the world. World production data for the years 1913 to the most recent represent the summed weights of all talc, pyrophyllite, soapstone, steatite, and other unspecified talc-related materials that were produced annually throughout the world." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹⁰² Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Talc or Pyrophyllite were used as keywords found in the patent title or abstract by year of publication.

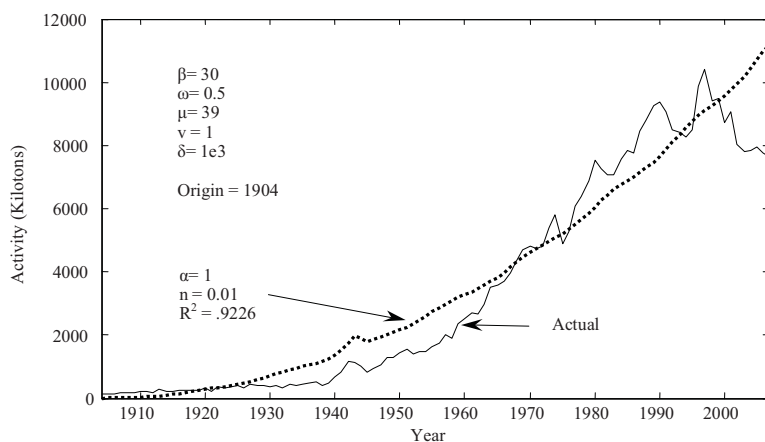


Figure A3.192. USGS World Talc Production. World talc and pyrophyllite production (activity) scaled in metric kilotons with actual and best-fit curves and common pattern equation parameters.

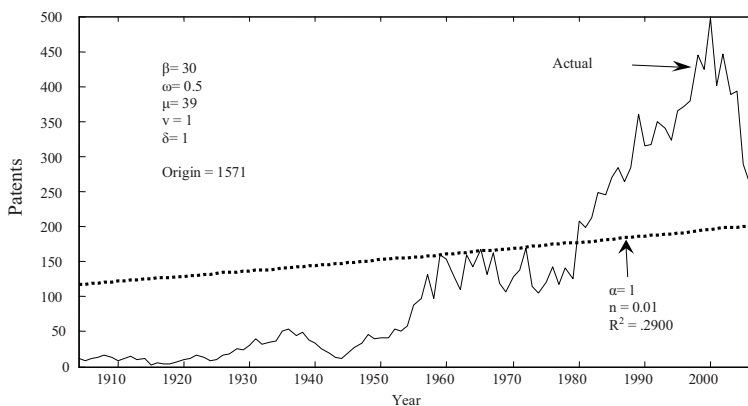


Figure A3.193. EPO Worldwide Patent Search: Talc or Pyrophyllite in Title or Abstract by date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

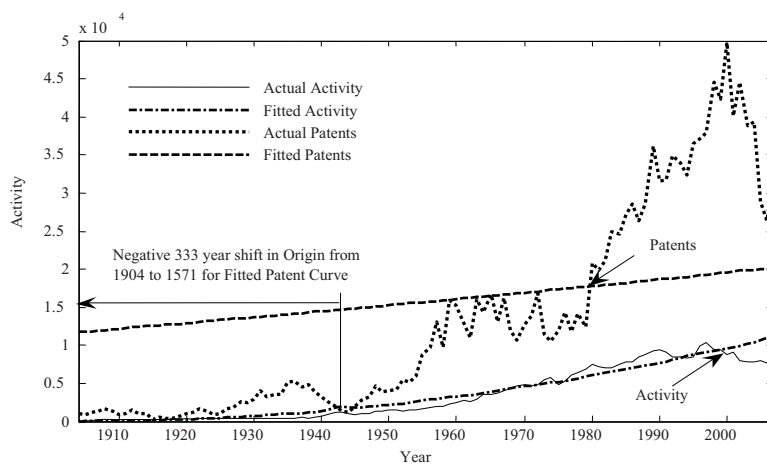


Figure A3.194. Talc and Pyrophyllite Best-Fit Patents and Activity. Illustrates best-fit origin shift.

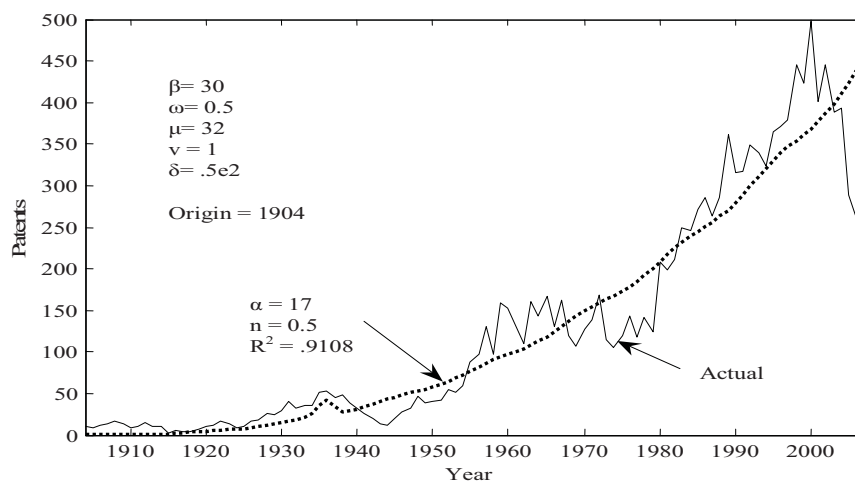


Figure A3.195. Talc and Pyrophyllite Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.87 Tantalum Activity¹⁰³ and Patents¹⁰⁴

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981	403	614
1901			1928			1955			1982	284	801
1902			1929			1956			1983	313	877
1903			1930			1957			1984	315	883
1904			1931			1958			1985	315	1031
1905			1932			1959			1986	215	1126
1906			1933			1960			1987	275	1187
1907			1934			1961			1988	292	1245
1908			1935			1962			1989	395	1561
1909			1936			1963			1990	396	1491
1910			1937			1964			1991	477	1577
1911			1938			1965			1992	399	1775
1912			1939			1966			1993	292	1557
1913			1940			1967			1994	333	1587
1914			1941			1968			1995	361	1436
1915			1942			1969	388	319	1996	436	1402
1916			1943			1970	318	360	1997	562	1468
1917			1944			1971	496	368	1998	779	1674
1918			1945			1972	371	412	1999	656	1802
1919			1946			1973	384	357	2000	1070	2210
1920			1947			1974	436	328	2001	1180	2304
1921			1948			1975	411	351	2002	1340	2513
1922			1949			1976	339	381	2003	1390	2370
1923			1950			1977	409	393	2004	1520	2241
1924			1951			1978	362	422	2005	1470	2021
1925			1952			1979	476	410	2006	964	2030
1926			1953			1980	544	584	2007	815	2098

¹⁰³ Activity represents world production of talc, defined at usgs.gov as "...data for the years 1964–68 were not available. World production data for the years 1969 to the most recent represent the tantalum content in tantalum-bearing ores and mineral concentrates that were produced from mines throughout the world. World production data for the years 1969 to the most recent were recorded from the MYB [*Minerals Yearbook*].” Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹⁰⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Talc or Pyrophyllite were used as keywords found in the patent title or abstract by year of publication.

Table A3.88 Correlation Eq.(A1.1) terms calculated from Table A3.87 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
22181	47566	17723137	76964848	34118459	5107835.4	18951403	7065601	0.718141	51.57272

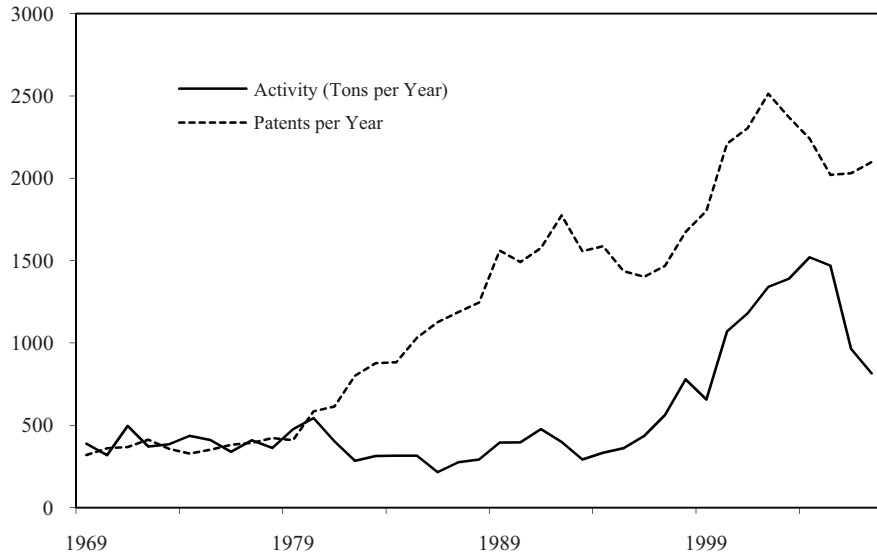


Figure A3.196. Tantalum Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

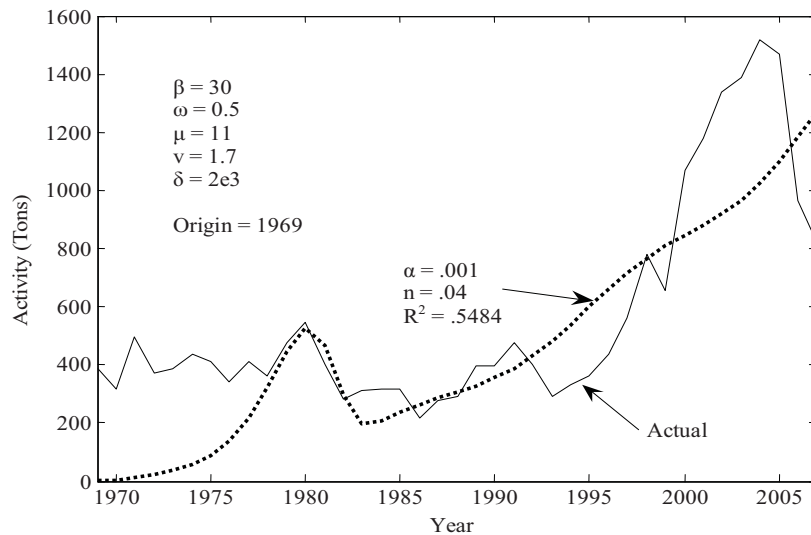


Figure A3.197. USGS World Tantalum Production. World tantalum production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

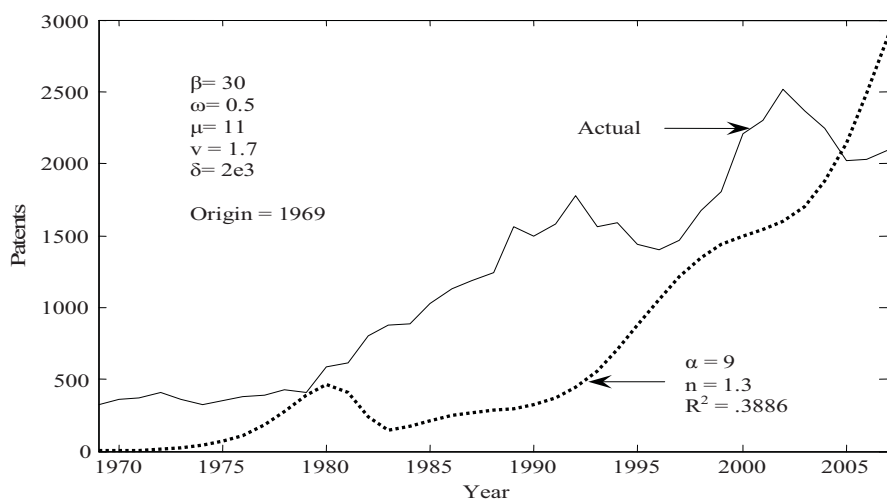


Figure A3.198. Tantalum Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.89 Tin Activity¹⁰⁵ and Patents¹⁰⁶

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	161000	180	1954	192000	255	1981	238000	1652
1901			1928	180000	183	1955	200000	268	1982	219000	1873
1902			1929	196000	195	1956	203000	284	1983	197000	2096
1903			1930	179000	200	1957	204000	303	1984	188000	2154
1904			1931	149000	198	1958	156000	277	1985	181000	2345
1905	93600	117	1932	96500	210	1959	164000	309	1986	173000	2583
1906	98400	95	1933	90400	187	1960	183000	452	1987	180000	2742
1907	93800	104	1934	122000	159	1961	187000	378	1988	205000	2803
1908	106000	107	1935	137000	209	1962	190000	471	1989	233000	3165
1909	106000	118	1936	182000	223	1963	194000	517	1990	221000	3097
1910	105000	116	1937	213000	207	1964	197000	542	1991	201000	2965
1911	112000	120	1938	166000	200	1965	204000	650	1992	191000	3382
1912	122000	121	1939	180000	181	1966	211000	604	1993	190000	3078
1913	136000	107	1940	240000	168	1967	218000	705	1994	178000	3149
1914	128000	100	1941	244000	154	1968	232000	680	1995	201000	3009
1915	129000	76	1942	124000	108	1969	229000	632	1996	220000	2987
1916	128000	75	1943	146000	100	1970	232000	731	1997	241000	2929
1917	135000	58	1944	102000	75	1971	235000	796	1998	231000	3259
1918	128000	60	1945	88400	99	1972	244000	954	1999	245000	3459
1919	123000	81	1946	89400	126	1973	238000	757	2000	278000	4068
1920	126000	109	1947	115000	151	1974	233000	697	2001	246000	4031
1921	110000	134	1948	156000	197	1975	222000	842	2002	233000	4173
1922	127000	150	1949	164000	170	1976	218000	945	2003	258000	4010
1923	130000	152	1950	172000	132	1977	231000	1002	2004	298000	3945
1924	142000	150	1951	172000	137	1978	241000	1069	2005	292000	3569
1925	147000	176	1952	177000	210	1979	245000	1134	2006	302000	3760
1926	146000	140	1953	193000	176	1980	245000	1640	2007	320000	3697

¹⁰⁵ Activity represents world production of talc, defined at usgs.gov as "...tin content of mine and mill production. Data were from the MYB [Minerals Yearbook] and MR [Mineral Resources of the United States]. Blank cells in the worksheet indicate that data were not available for the years 1900–04." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹⁰⁶ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Talc or Pyrophyllite were used as keywords found in the patent title or abstract by year of publication.

Table A3.90 Correlation Eq.(A1.1) terms calculated from Table A3.89 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
18814500	108875	3.73E+12	2.89E+08	2.46E+10	2.959E+11	1.74E+08	4.72E+09	0.657578	43.24093

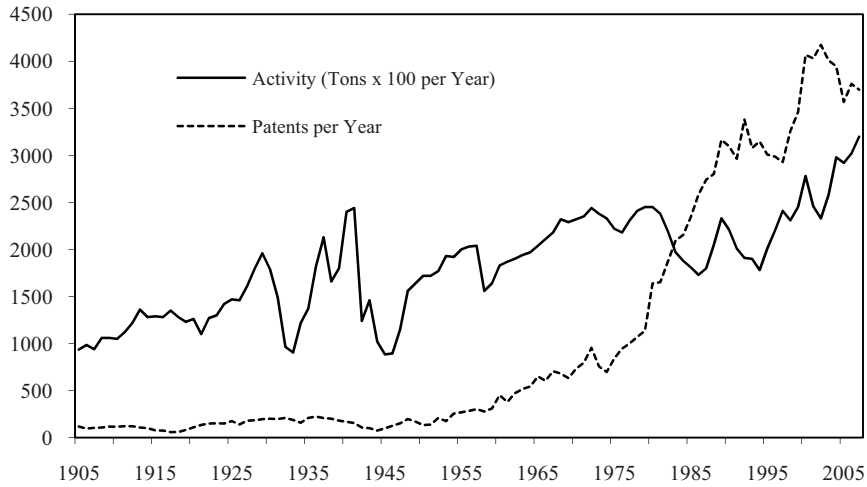


Figure A3.199. Tin Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

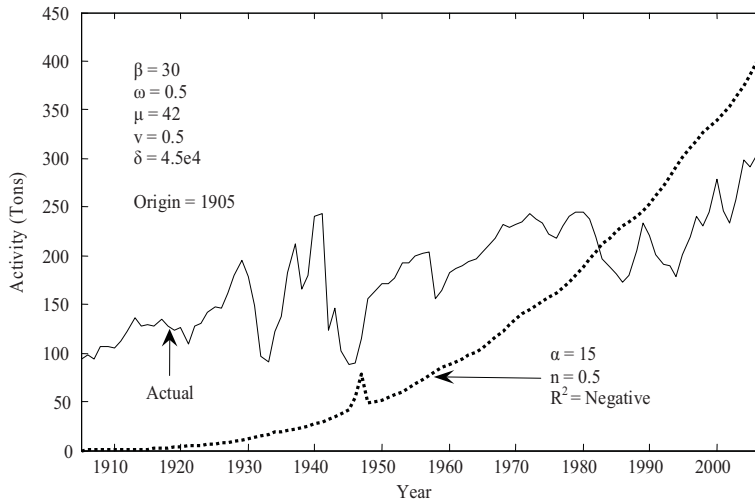


Figure A3.200. USGS World Tin Production. World tin production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters. The negative R^2 may indicate possible Stage IV. No best-fit for the patent data was obtainable.

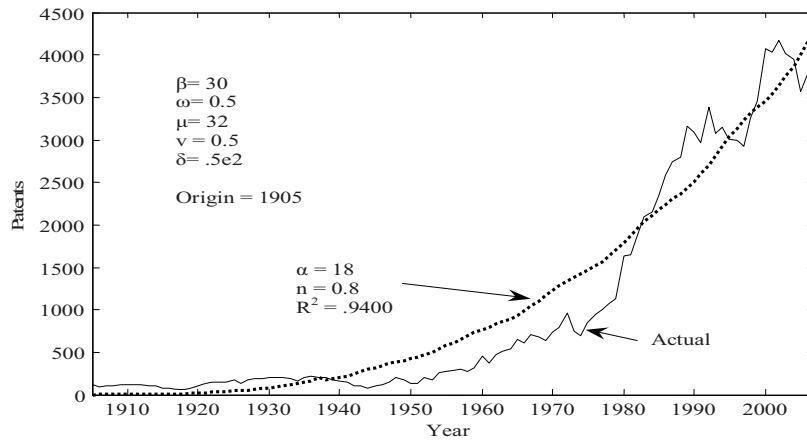


Figure A3.201. Tin Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.91 Titanium Activity¹⁰⁷ and Patents¹⁰⁸

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	36100	60	1954	1180000	317	1981	5140000	2623
1901			1928	40100	80	1955	1340000	450	1982	4420000	3105
1902			1929	48200	91	1956	1750000	493	1983	4040000	3283
1903			1930	43500	123	1957	1930000	569	1984	5320000	3469
1904			1931	43700	150	1958	1650000	650	1985	5110000	3869
1905			1932	65700	150	1959	1830000	707	1986	5100000	4333
1906			1933	79000	133	1960	2100000	984	1987	5950000	4676
1907			1934	106000	152	1961	2230000	836	1988	6280000	5038
1908			1935	174000	154	1962	2100000	835	1989	6570000	5696
1909			1936	226000	192	1963	2190000	856	1990	6250000	5769
1910			1937	284000	210	1964	2540000	828	1991	5330000	5868
1911			1938	323000	226	1965	2680000	971	1992	6050000	6557
1912			1939	89600	208	1966	2870000	786	1993	6040000	6116
1913			1940	361000	164	1967	3020000	1007	1994	6030000	6361
1914			1941	248000	170	1968	3230000	952	1995	6240000	6030
1915			1942	217000	156	1969	3610000	879	1996	6210000	6148
1916			1943	379000	123	1970	4020000	1064	1997	6450000	6344
1917			1944	483000	109	1971	3750000	1084	1998	7050000	8456
1918			1945	532000	108	1972	3610000	1340	1999	6550000	8690
1919			1946	527000	132	1973	3920000	1221	2000	7350000	9308
1920			1947	704000	146	1974	4400000	1056	2001	7570000	8650
1921			1948	736000	190	1975	4030000	1308	2002	7730000	8938
1922			1949	844000	205	1976	4390000	1290	2003	8210000	8816
1923			1950	884000	186	1977	4350000	1519	2004	8380000	9298
1924			1951	160000	199	1978	4760000	1646	2005	8450000	8326
1925	14800	49	1952	942000	288	1979	4660000	1611	2006	9740000	8165
1926	21400	47	1953	860000	228	1980	5380000	2390	2007	10000000	8520

¹⁰⁷ Activity represents world production of titanium, defined at usgs.gov as "...ilmenite and natural rutile, and titanium slag, but does not include ilmenite used to produce titanium slag to avoid double counting. Data are not available prior to 1925. Titanium slag was not produced prior to 1950. Data are from the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]; the typical silicon content of silicon metal is 98% of the gross weight." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹⁰⁸ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Titanium or Ti were used as keywords found in the patent title or abstract by year of publication.

Table A3.92. Correlation Eq.(A1.1) terms calculated from Table A3.91 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
270552100	204530	1.52E+15	1.25E+09	1.3E+12	6.429E+14	7.47E+08	6.34E+11	0.915079	83.73694

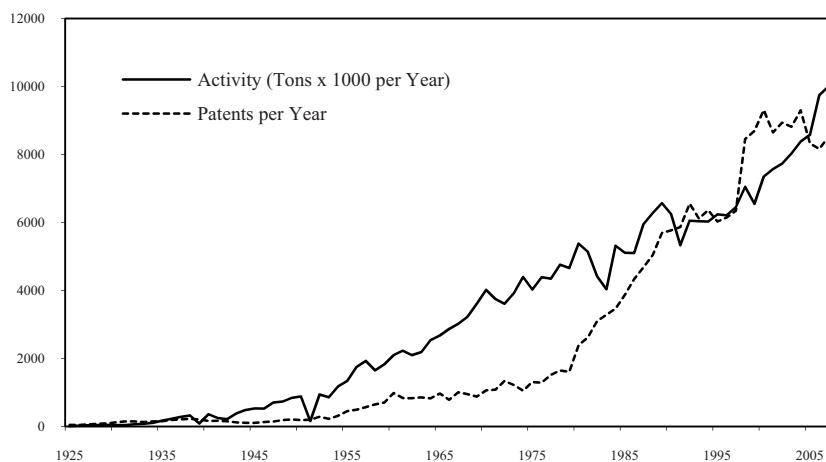


Figure A3.202. Titanium Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

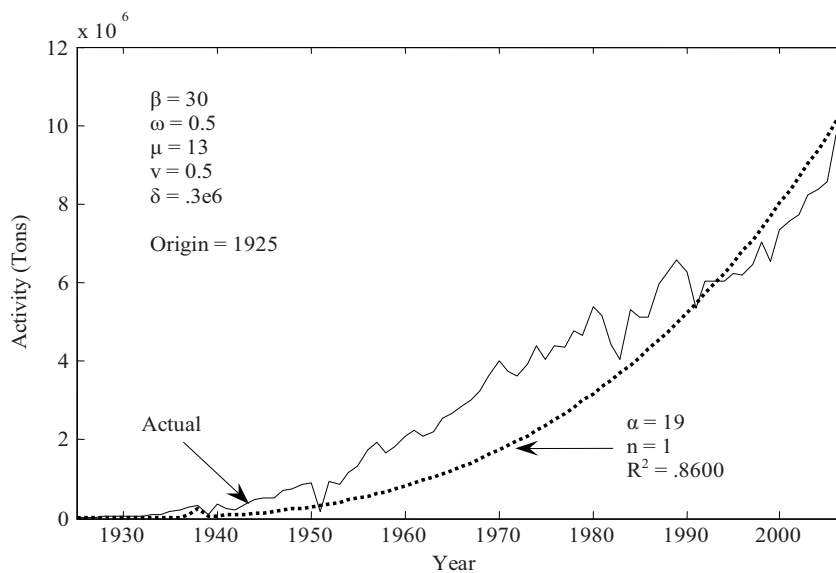


Figure A3.203. USGS World Titanium Production. World titanium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

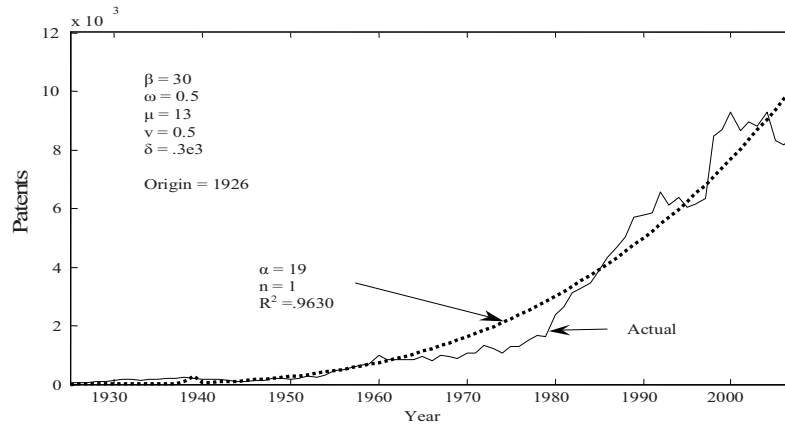


Figure A3.204. EPO Worldwide Patent Search: Titanium or Ti in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

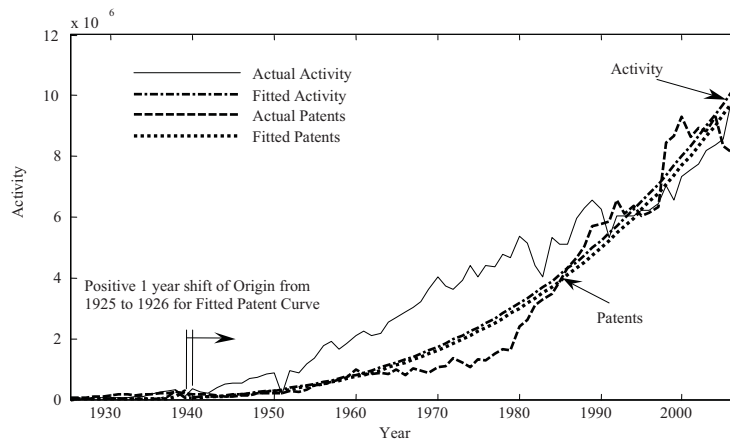


Figure A3.205. Titanium Best-Fit Activity and Patents. Illustrates best-fit origin shift.

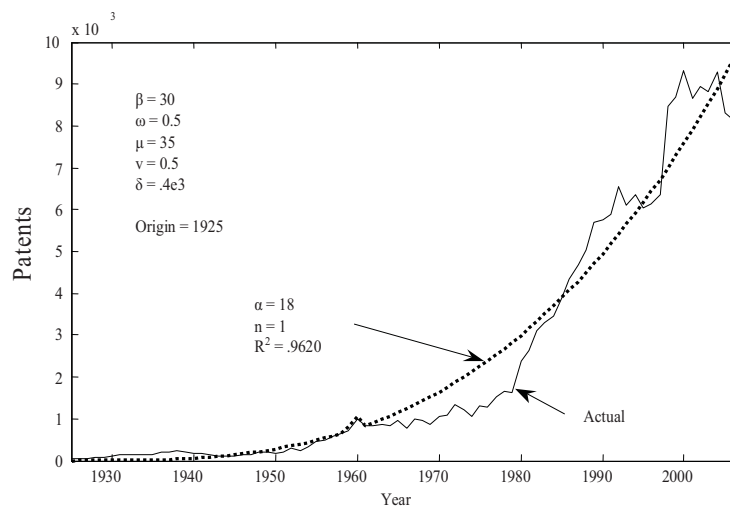


Figure A3.206. Titanium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.93 Tungsten Activity¹⁰⁹ and Patents¹¹⁰

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927	4400	100	1954	33800	160	1981	50300	602
1901			1928	5500	144	1955	35700	205	1982	47000	666
1902			1929	7500	145	1956	35800	169	1983	40900	740
1903			1930	7900	190	1957	29100	233	1984	46200	787
1904			1931	6400	244	1958	24200	200	1985	46600	890
1905	1700	14	1932	3200	230	1959	26400	195	1986	43500	1036
1906	1900	22	1933	5900	193	1960	31200	293	1987	42500	1030
1907	2600	55	1934	7800	187	1961	33000	260	1988	50900	1180
1908	1800	64	1935	10700	187	1962	31300	240	1989	51000	1331
1909	2500	39	1936	11800	189	1963	27100	289	1990	51900	1409
1910	3300	34	1937	18500	213	1964	28100	327	1991	48200	1323
1911	3200	48	1938	17800	245	1965	27000	341	1992	42900	1581
1912	4200	35	1939	20100	194	1966	28600	297	1993	34300	1285
1913	3900	65	1940	20700	126	1967	28500	380	1994	34000	1307
1914	3500	46	1941	23900	123	1968	31000	398	1995	38500	1201
1915	5200	41	1942	24100	85	1969	32500	380	1996	34700	1288
1916	10000	33	1943	28600	75	1970	32400	398	1997	33200	1230
1917	12300	26	1944	23400	73	1971	35400	417	1998	37000	1523
1918	15200	29	1945	10900	89	1972	38500	511	1999	37700	1616
1919	7000	31	1946	9040	131	1973	37900	501	2000	44000	2237
1920	5500	49	1947	13700	122	1974	37600	429	2001	50800	2217
1921	2300	63	1948	17800	139	1975	38300	419	2002	47000	2341
1922	3000	75	1949	15800	152	1976	38000	441	2003	47200	2094
1923	3300	87	1950	18300	117	1977	41100	411	2004	66600	1955
1924	2900	94	1951	24800	124	1978	46100	424	2005	59600	1584
1925	4900	123	1952	32700	174	1979	48600	415	2006	56600	1559
1926	5800	97	1953	34400	128	1980	52000	559	2007	54500	1603

Table A3.94. Correlation Eq.(A1.1) terms calculated from Table A3.93 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
2698440	52196	1.01E+11	62604270	2.15E+09	3.05E+10	36153567	7.79E+08	0.741966	55.05131

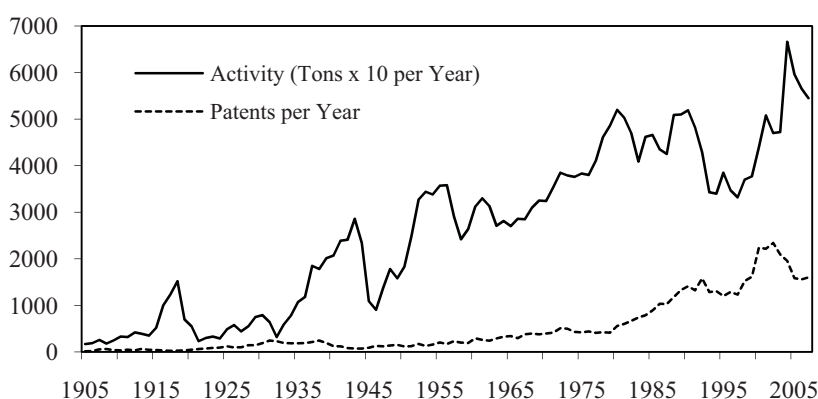


Figure A3.207. Tungsten Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

¹⁰⁹ Activity represents world production of tungsten, defined at usgs.gov as "...tungsten content of concentrate. Data for the years 1905–2001 were from the MR [*Mineral Resources of the United States*] and the MYB [*Minerals Yearbook*]; datum for 2002 is a previously unpublished revision; data for 2003 to the most recent year are from the MYB. Blank cells in the worksheet indicate that data were not available for the years 1900–04." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹¹⁰ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Tungsten or Wolfram were used as keywords found in the patent title or abstract by year of publication.

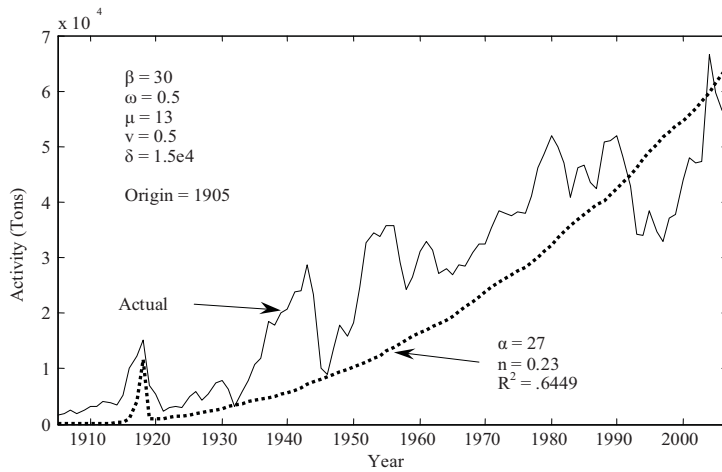


Figure A3.208. USGS World Tungsten Production. World tungsten production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

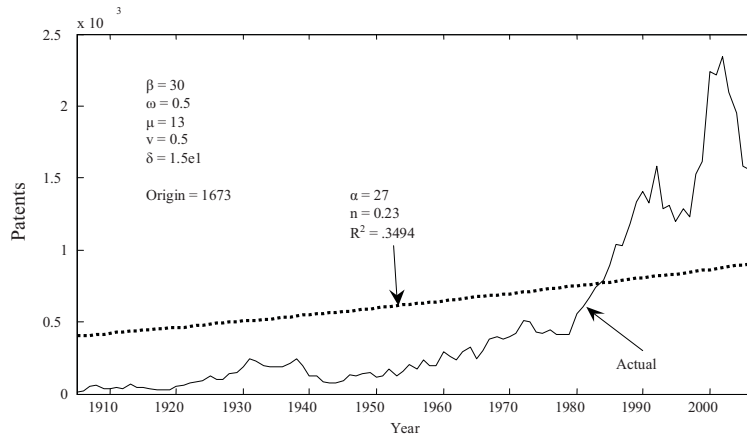


Figure A3.209. EPO Worldwide Patent Search: Tungsten or Wolfram in Title or Abstract by date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

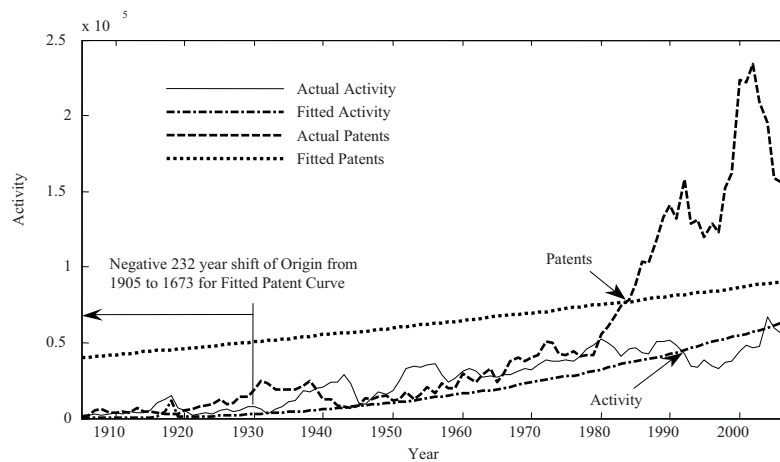


Figure A3.210. Tungsten Best-Fit Activity and Patents. Illustrates best-fit origin shift.

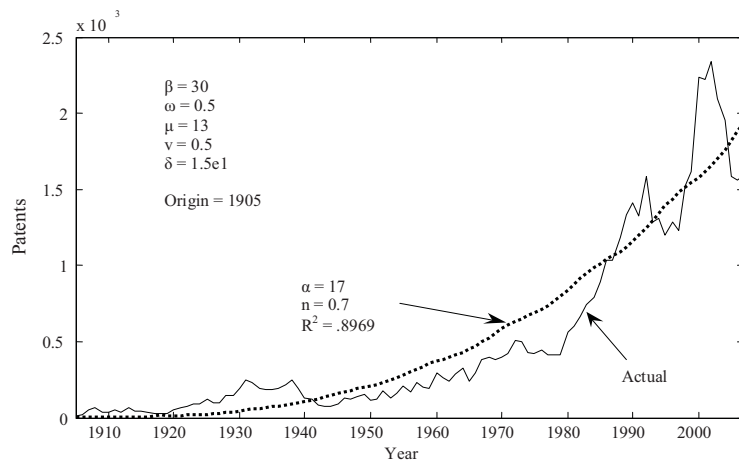


Figure A3.211. Tungsten Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.95 Vanadium Activity¹¹¹ and Patents¹¹²

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981	35300	501
1901			1928			1955			1982	27200	500
1902			1929			1956			1983	27200	447
1903			1930			1957			1984	31100	501
1904			1931			1958			1985	31000	475
1905			1932			1959			1986	32000	415
1906			1933			1960	5040	196	1987	32000	476
1907			1934			1961	7850	154	1988	33000	471
1908			1935			1962	6080	164	1989	33000	510
1909			1936			1963	6500	160	1990	33200	503
1910			1937			1964	7170	172	1991	26400	553
1911			1938			1965	8300	215	1992	26700	624
1912			1939			1966	8440	173	1993	2500	607
1913			1940			1967	9610	217	1994	3200	616
1914			1941			1968	11400	216	1995	3600	542
1915			1942			1969	10300	172	1996	35100	587
1916			1943			1970	14900	206	1997	37100	599
1917			1944			1971	15800	257	1998	42700	736
1918			1945			1972	15500	250	1999	36300	764
1919			1946			1973	16000	215	2000	41000	949
1920			1947			1974	20400	203	2001	41800	877
1921			1948			1975	21600	292	2002	51000	880
1922			1949			1976	29200	288	2003	47900	1017
1923			1950			1977	29000	326	2004	51900	899
1924			1951			1978	29400	409	2005	56400	780
1925			1952			1979	37700	329	2006	56300	840
1926			1953			1980	35900	454	2007	58500	1027

Activity represents world production of Vanadium, defined at usgs.gov as "...mine production of vanadium. Data were from the MR [*Mineral Resources of the United States*] and MYB [*Minerals Yearbook*] for the years 1912–22, 1925, 1927–31, 1934–43, 1945–47, and 1998 to the most recent, the CDS [*Commodity Data Summaries*] for the years 1960–77, and the MCS for the years 1978–84 and 1990–97. Blank cells in the worksheet indicate that data were not available for the years 1900–11, 1923–24, and 1948–59. World production was interpolated to two significant figures for the years 1926, 1932–33, 1944, and 1985–89. World production data for the years 1927–31 and 1997–99 do not contain U.S. production." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹¹² Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Vanadium was used as the keyword found in the patent title or abstract by year of publication.

Table A3.96. Correlation Eq.(A1.1) terms calculated from Table A3.95 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
1279490	22764	4.58E+10	13890388	7.56E+08	1.166E+10	3094561	1.49E+08	0.783991	61.46415

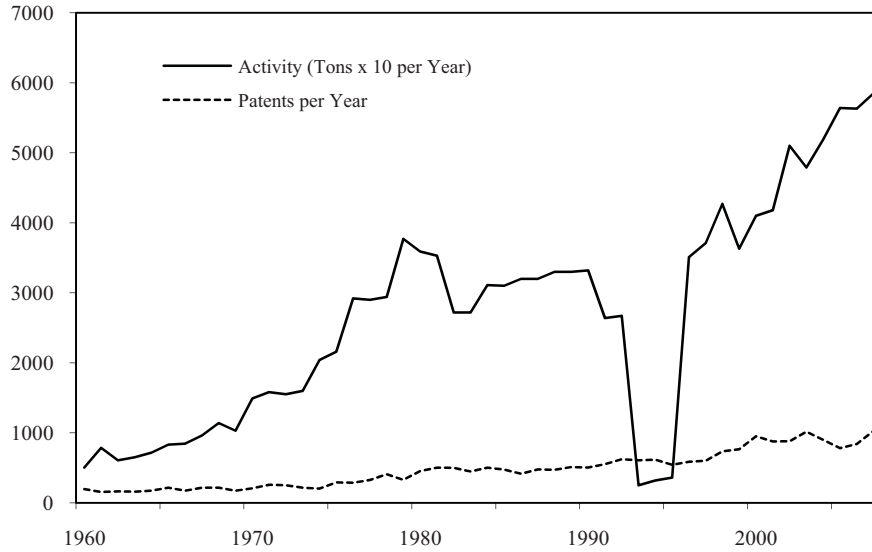


Figure A3.212. Vanadium Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

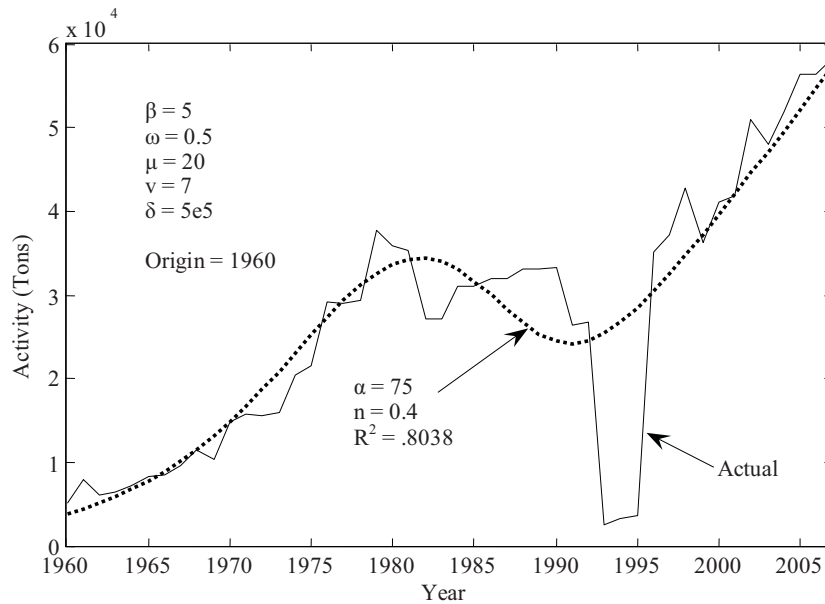


Figure A3.213. USGS World Vanadium Production. World vanadium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

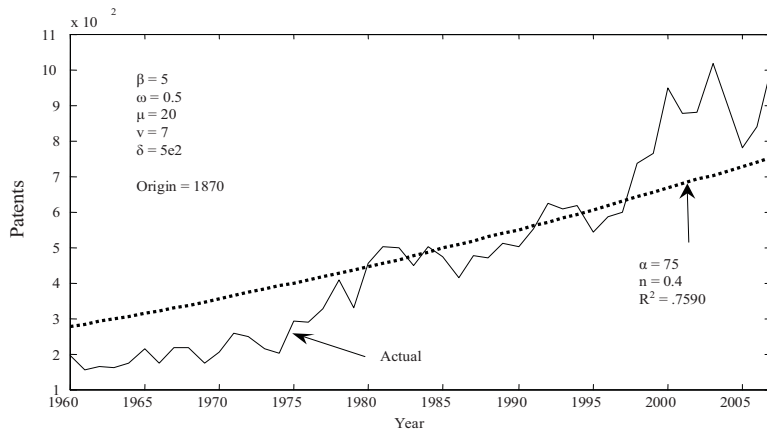


Figure A3.214. EPO Worldwide Patent search: Vanadium in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

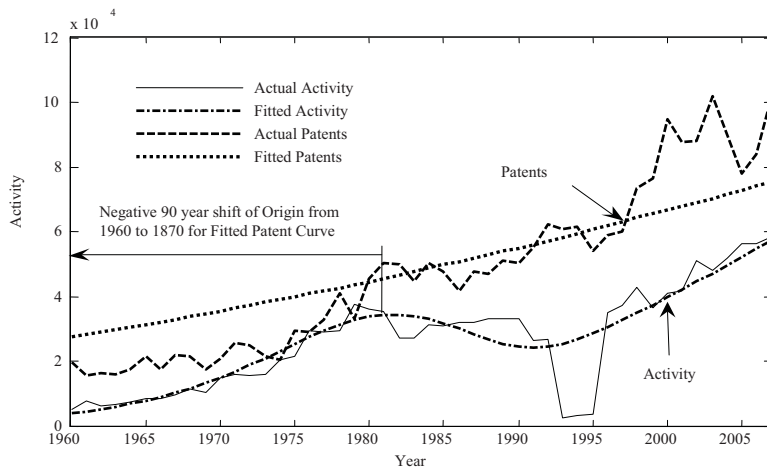


Figure A3.215. Vanadium Best-Fit Activity and Patents. Illustrates best-fit origin shift.

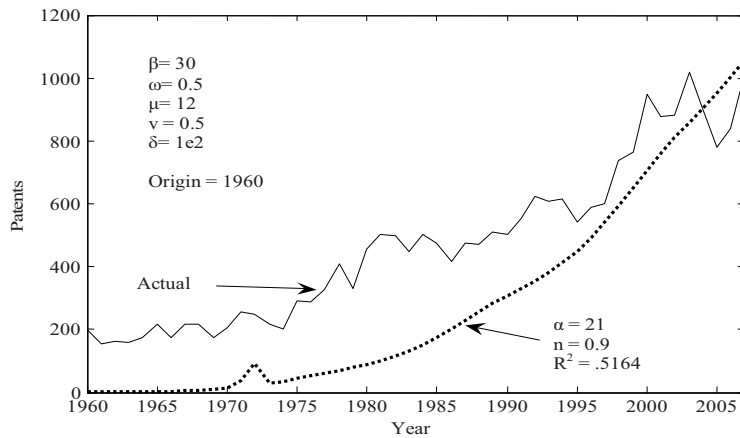


Figure A3.216. Vanadium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.97 Zinc Activity¹¹³ and Patents¹¹⁴

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	479000	141	1927	1420000	313	1954	2660000	647	1981	5950000	2788
1901	510000	121	1928	1360000	361	1955	2900000	699	1982	6130000	3144
1902	547000	120	1929	1320000	384	1956	3110000	796	1983	6280000	3144
1903	574000	156	1930	1260000	517	1957	3150000	781	1984	6520000	3247
1904	629000	134	1931	904000	556	1958	2950000	715	1985	6760000	3367
1905	660000	122	1932	709000	488	1959	3020000	752	1986	6840000	3682
1906	704000	135	1933	892000	506	1960	3090000	1119	1987	7190000	3627
1907	738000	146	1934	1060000	511	1961	3490000	1019	1988	6770000	4000
1908	723000	151	1935	1210000	549	1962	3570000	924	1989	6820000	4387
1909	775000	132	1936	1330000	542	1963	3660000	992	1990	7150000	4233
1910	810000	148	1937	1470000	496	1964	4030000	1022	1991	7270000	4395
1911	895000	144	1938	1420000	643	1965	4310000	1164	1992	7250000	4914
1912	971000	154	1939	1500000	470	1966	4500000	1042	1993	6910000	4485
1913	939000	142	1940	1470000	357	1967	4840000	1291	1994	7050000	4561
1914	795000	115	1941	1590000	320	1968	4970000	1167	1995	7280000	4277
1915	760000	101	1942	1630000	274	1969	5340000	1087	1996	7480000	4352
1916	882000	86	1943	1830000	225	1970	5460000	1231	1997	7540000	4157
1917	901000	77	1944	1870000	191	1971	5520000	1335	1998	7570000	4851
1918	849000	74	1945	1470000	306	1972	5440000	1485	1999	7960000	4908
1919	719000	124	1946	1440000	327	1973	5710000	1339	2000	8770000	5522
1920	682000	162	1947	1600000	365	1974	5780000	1217	2001	8910000	5554
1921	464000	234	1948	1690000	492	1975	5850000	1510	2002	8880000	5953
1922	730000	228	1949	1730000	468	1976	5690000	1610	2003	9520000	5954
1923	889000	222	1950	2150000	385	1977	5920000	1677	2004	9590000	5797
1924	986000	225	1951	2360000	503	1978	5850000	2034	2005	9930000	5689
1925	1190000	261	1952	2590000	608	1979	5990000	2066	2006	10000000	5645
1926	1410000	254	1953	2670000	482	1980	5950000	2698	2007	10900000	5884

Table A3.98. Correlation Eq.(A1.1) terms calculated from Table A3.97 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
395126000	171684	2.35E+15	6.34E+08	1.16E+12	9.058E+14	3.61E+08	5.37E+11	0.938743	88.12379

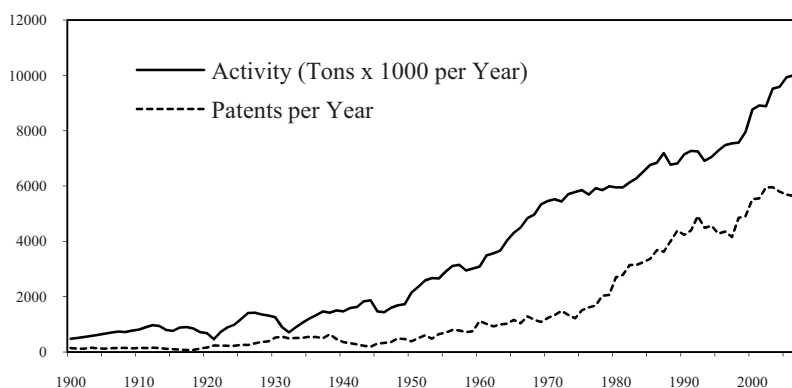


Figure A3.217. Zinc Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

Activity represents world production of Zinc, defined at usgs.gov as "...zinc content of smelter production for the years 1900–12, 1914–17, and 1929–42. World mine production data were used for the years 1913, 1918–28, and 1943 to the most recent. Data were from the MR [*Mineral Resources of the United States*] and MYB [*Minerals Yearbook*]." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹¹⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Zinc or Zn were used as keywords found in the patent title or abstract by year of publication.

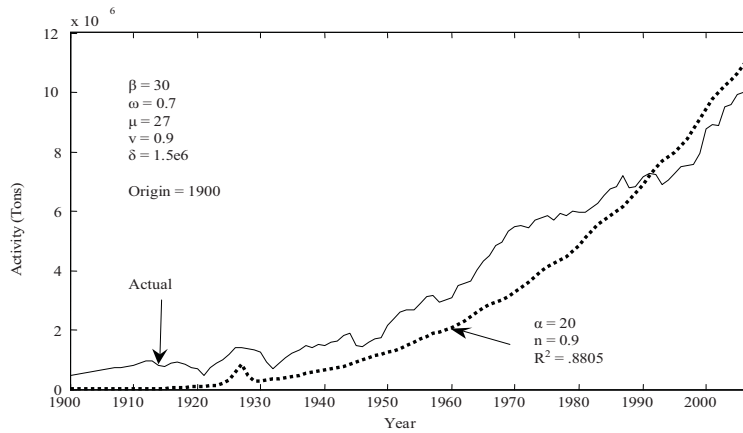


Figure A3.218. USGS World Zinc Production. World zinc production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

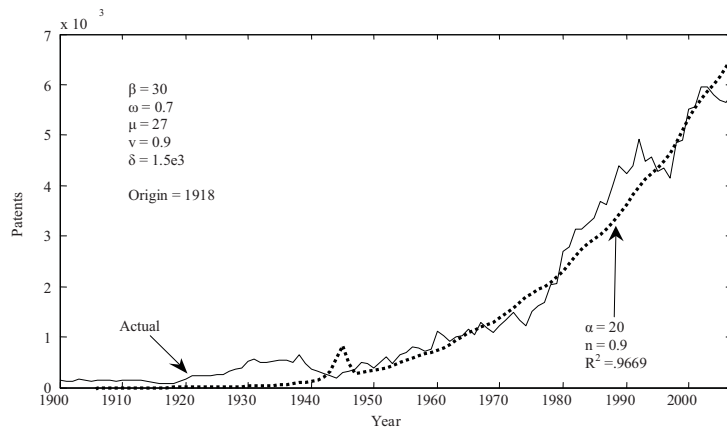


Figure A3.219. EPO Worldwide Patent Search: Zinc or Zn in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

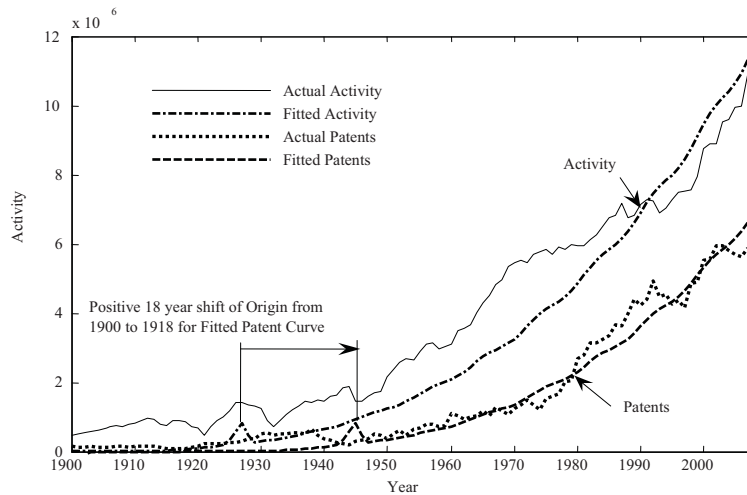


Figure A3.220. Zinc Best-Fit Activity and Patents. Illustrates best-fit origin shift.

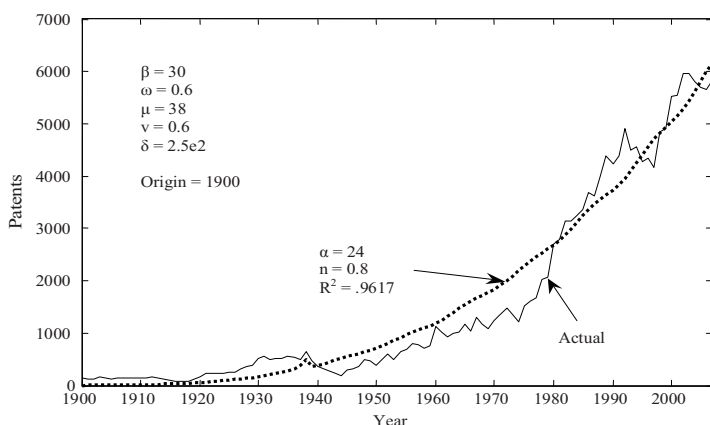


Figure A3.221. Zinc Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A3.99 Zirconium Activity¹¹⁵ and Patents¹¹⁶

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	63200	109	1981	645000	908
1901			1928			1955	78200	170	1982	710000	1030
1902			1929			1956	118000	186	1983	666000	1204
1903			1930			1957	146000	256	1984	736000	1308
1904			1931			1958	105000	270	1985	815000	1419
1905			1932			1959	103000	272	1986	741000	1596
1906			1933			1960	129000	417	1987	753000	1777
1907			1934			1961	159000	371	1988	929000	1941
1908			1935			1962	149000	387	1989	979000	2259
1909			1936			1963	195000	365	1990	852000	2155
1910			1937			1964	191000	363	1991	795000	2173
1911			1938			1965	235000	408	1992	856000	2379
1912			1939			1966	244000	383	1993	796000	2292
1913			1940			1967	293000	475	1994	897000	2257
1914			1941			1968	309000	476	1995	918000	2165
1915			1942			1969	386000	405	1996	894000	2083
1916			1943			1970	399000	478	1997	830000	2004
1917			1944	17300	37	1971	432000	443	1998	732000	2419
1918			1945	19700	48	1972	369000	557	1999	673000	2503
1919			1946	24700	51	1973	379000	497	2000	731000	2711
1920			1947	25900	53	1974	397000	445	2001	750000	2604
1921			1948	26600	77	1975	418000	535	2002	973000	2960
1922			1949	24100	88	1976	448000	503	2003	1030000	2841
1923			1950	25200	73	1977	505000	562	2004	1090000	3106
1924			1951	46200	126	1978	525000	658	2005	1100000	2694
1925			1952	33900	114	1979	629000	623	2006	1250000	2879
1926			1953	53300	80	1980	680000	867	2007	1470000	3082

Activity represents world production of Zirconium, defined at usgs.gov as "...zirconium mineral concentrates. Data were from the MR [*Mineral Resources of the United States*] and MYB [*Minerals Yearbook*]. Blank cells in the worksheet indicate that data were not available for the years 1900–43. Production data for the United States were not included for the years 1944–52, 1959–87, and 1993 to the most recent." Data is in metric tons as reported by the United States Geologic Survey (USGS) at minerals.usgs.gov.

¹¹⁶ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Zirconium or Zr were used as keywords found in the patent title or abstract by year of publication.

Table A3.100. Correlation Eq.(A1.1) terms calculated from Table A3.99 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
31992300	70977	2.48E+13	1.43E+08	5.75E+10	8.797E+12	64722075	2.2E+10	0.923869	85.35335

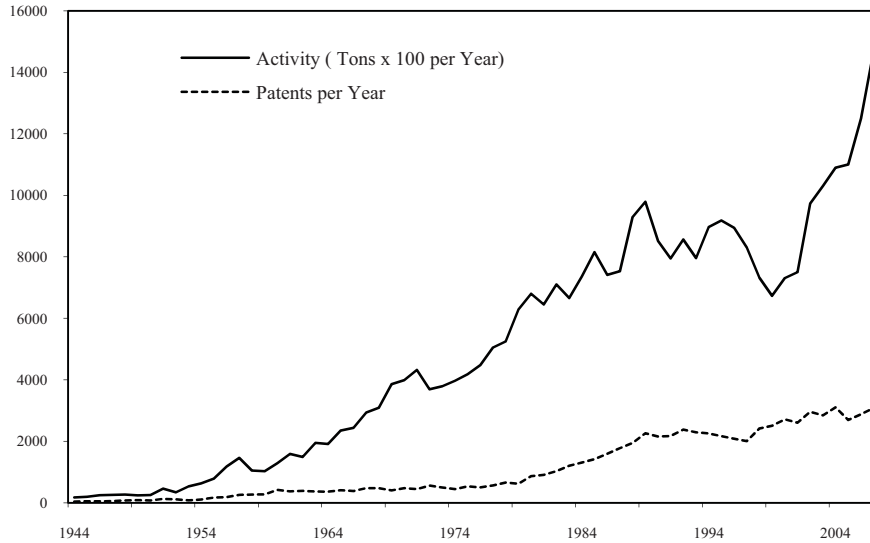


Figure A3.222. Zirconium Activity and Patents. data illustrates correlation. Activity scaled to fit plot.

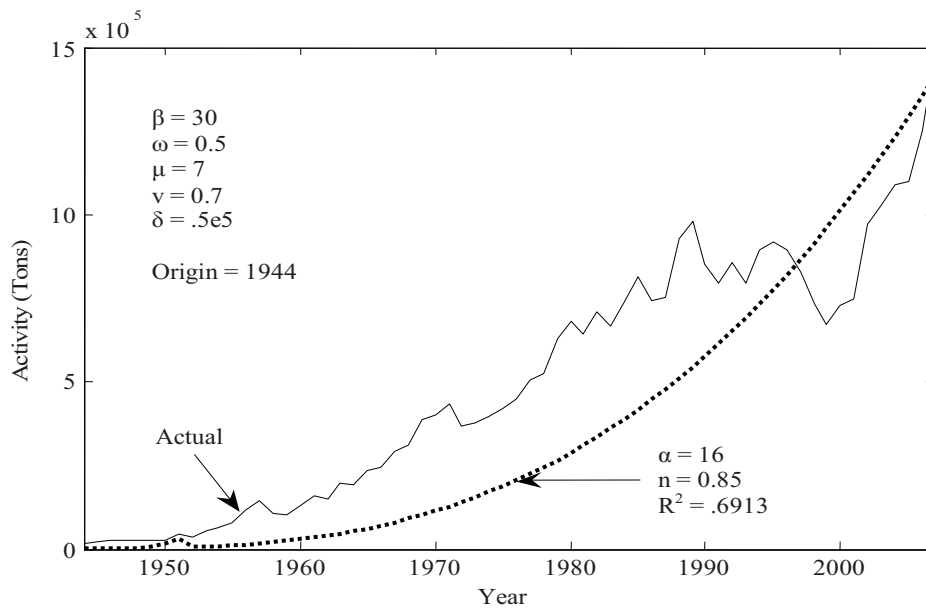


Figure A3.223. USGS World Zirconium Production. World zirconium production (activity) scaled in metric tons with actual and best-fit curves and common pattern equation parameters.

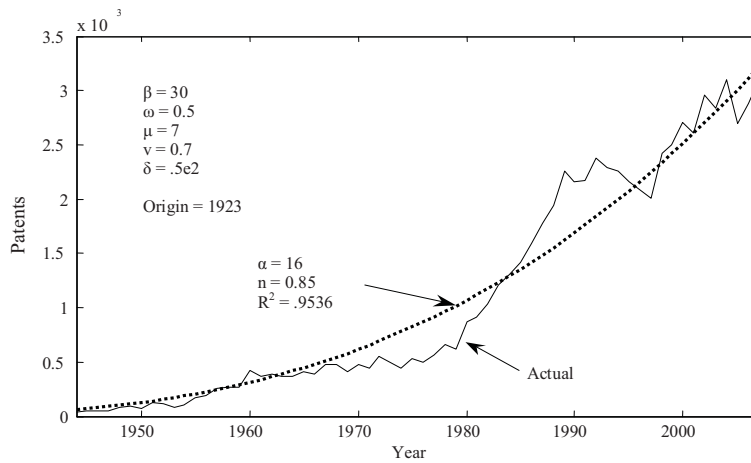


Figure A3.224. EPO Worldwide patent Search: Zirconium or Zr in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

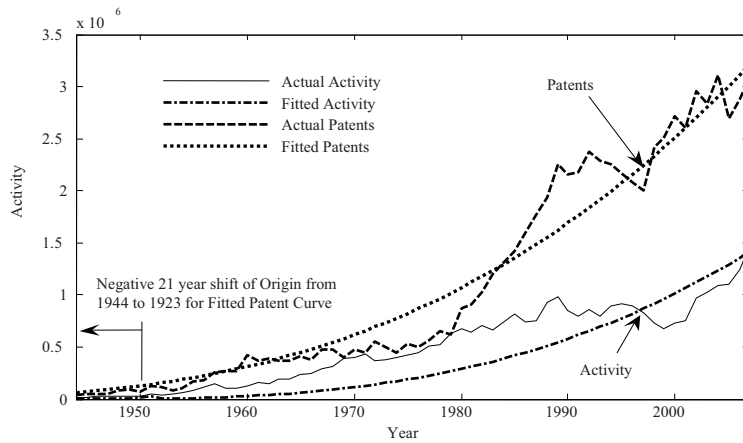


Figure A3.225. Zirconium Best-Fit Activity and Patents. Illustrates best-fit origin shift.

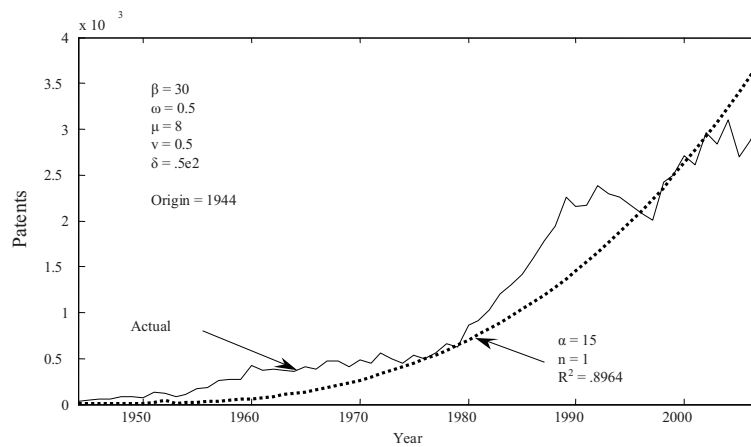


Figure A3.226. Zirconium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Appendix 4: Energy Sources Data

Table A4.1 U.S. Biofuel Energy Activity¹¹⁷ and Patents¹¹⁸

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981	13043.9	1
1901			1928			1955			1982	35281.53	1
1902			1929			1956			1983	64754.16	0
1903			1930			1957			1984	79274.4	1
1904			1931			1958			1985	95527.26	2
1905			1932			1959			1986	109831.4	0
1906			1933			1960			1987	125855.1	0
1907			1934			1961			1988	127221.9	0
1908			1935			1962			1989	128575.7	0
1909			1936			1963			1990	113694.6	1
1910			1937			1964			1991	131265.1	2
1911			1938			1965			1992	148704.8	0
1912			1939			1966			1993	173374.6	1
1913			1940			1967			1994	193197.2	8
1914			1941			1968			1995	204894.4	5
1915			1942			1969			1996	146431.5	5
1916			1943			1970			1997	188311.9	8
1917			1944			1971			1998	206162.7	10
1918			1945			1972			1999	214298.2	16
1919			1946			1973			2000	241725.6	12
1920			1947			1974			2001	259057.8	13
1921			1948			1975			2002	311003.3	15
1922			1949			1976			2003	415283.1	33
1923			1950			1977			2004	515133.9	62
1924			1951			1978			2005	596543.9	119
1925			1952			1979			2006	798629.3	167
1926			1953			1980			2007	1029982	426
									2008	1420147	638

Table A4.2 Correlation Eq.(A1.1) terms calculated from Table A4.1 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
8087206.9	1546	5.07E+12	636588	1.61E+09	2.733E+12	551226.7	1.16E+09	0.946938	89.66916

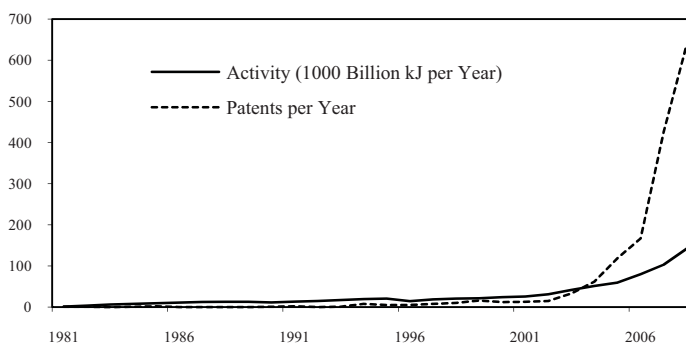


Figure A4.1. U.S. Biofuel Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

¹¹⁷ Activity represents United States production of biofuel energy, defined at eia.doe.gov as “energy from “total biomass inputs to the production of fuel ethanol and biodiesel.” Data is in billion Btu’s as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ.

¹¹⁸ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Biofuel or biofuels or biodiesel were used as keywords found in the patent title or abstract by year of publication.

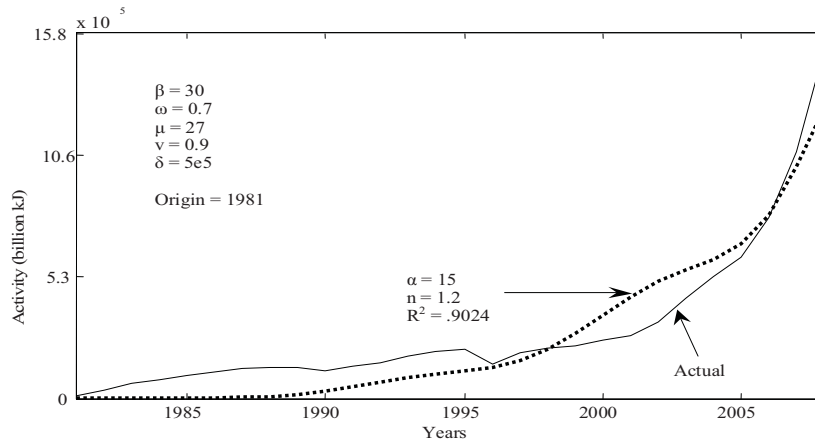


Figure A4.2. EIA U.S. Biofuel Energy Production. U.S. biofuel power production (activity) scaled in billion kJ with actual and best-fit curves and common pattern equation parameters.

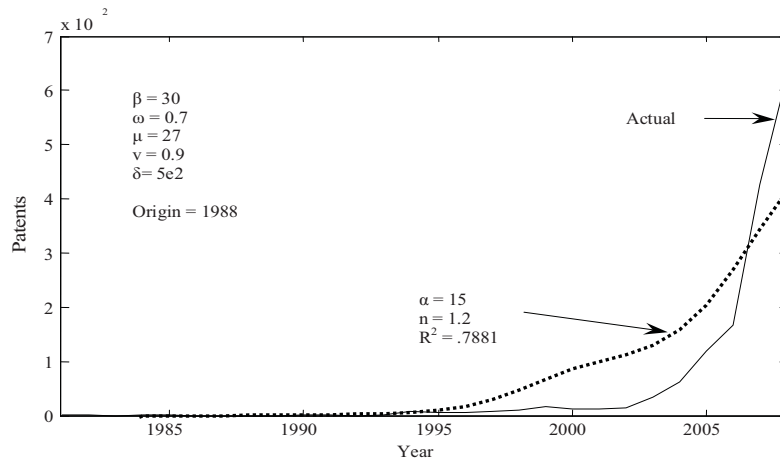


Figure A4.3. EPO Worldwide Patent Search: Biofuel or Biofuels or Biodiesel in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

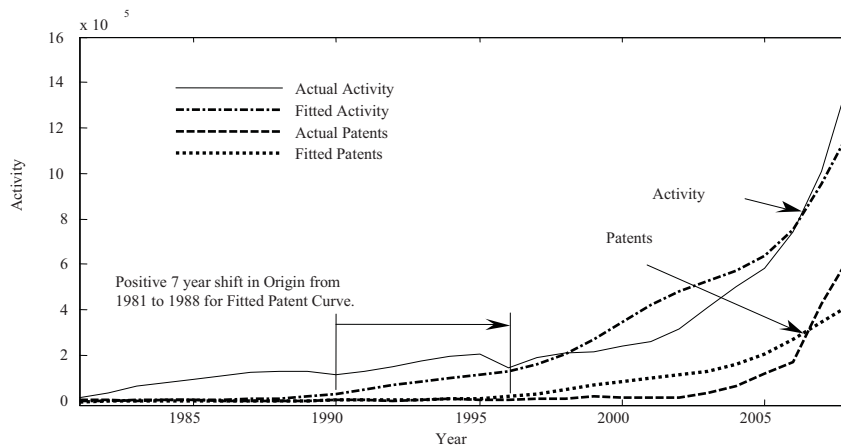


Figure A4.4. U.S. Biofuel Energy best-Fit Activity and Patents. Illustrates best-fit origin shift.

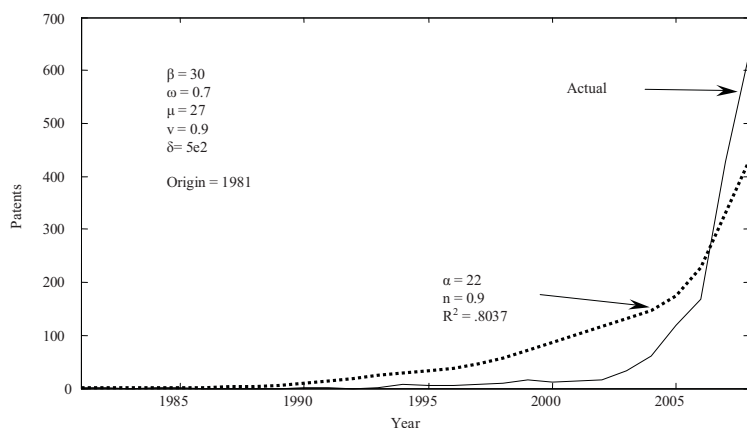


Figure A4.5. U.S. Biofuel Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.3 U.S. Biomass Energy Activity¹¹⁹ and Patents¹²⁰

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	1471015	0	1981	2739352	55
1901			1928			1955	1502471	0	1982	2810682	91
1902			1929			1956	1493744	0	1983	3065517	106
1903			1930			1957	1406928	0	1984	3136195	126
1904			1931			1958	1395895	0	1985	3184131	136
1905			1932			1959	1427282	0	1986	3095665	123
1906			1933			1960	1392463	0	1987	3035644	143
1907			1934			1961	1365974	0	1988	3184602	112
1908			1935			1962	1371755	0	1989	3335821	157
1909			1936			1963	1396098	0	1990	2887927	157
1910			1937			1964	1410326	0	1991	2937553	137
1911			1938			1965	1408173	0	1992	3096042	167
1912			1939			1966	1444279	0	1993	3072057	164
1913			1940			1967	1413963	0	1994	3198106	175
1914			1941			1968	1497567	0	1995	3273789	211
1915			1942			1969	1519714	0	1996	3331884	199
1916			1943			1970	1509665	0	1997	3282854	224
1917			1944			1971	1511101	0	1998	3094379	373
1918			1945			1972	1585734	7	1999	3132753	554
1919			1946			1973	1613167	7	2000	3175992	491
1920			1947			1974	1624338	1	2001	2773944	728
1921			1948			1975	1581164	10	2002	2860810	677
1922			1949	1634471	0	1976	1807609	13	2003	2969689	1912
1923			1950	1648234	0	1977	1939440	18	2004	3176138	1728
1924			1951	1619076	0	1978	2149673	41	2005	3291750	3179
1925			1952	1555459	0	1979	2270261	47	2006	3491022	1183
1926			1953	1496624	0	1980	2611653	52	2007	3780533	1503
									2008	4114410	2304

Activity represents U.S. production of biomass energy, defined at usgs.gov as energy from “Wood and wood derived fuels, biomass waste, fuel ethanol and biodiesel.” Data is in billion Btu’s as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ..

¹²⁰ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Biomass was used as the keyword found in the patent title or abstract by year of publication.

Table A4.4. Correlation Eq.(A1.1) terms calculated from Table A4.3 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
139604559	17311	3.66E+14	27790219	5.75E+10	4.132E+13	22795707	1.73E+10	0.562375	31.62656

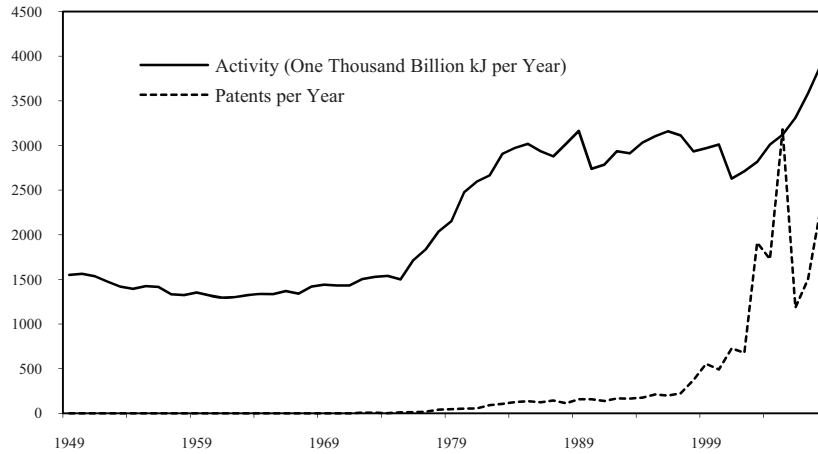


Figure A4.6. U.S. Biomass Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

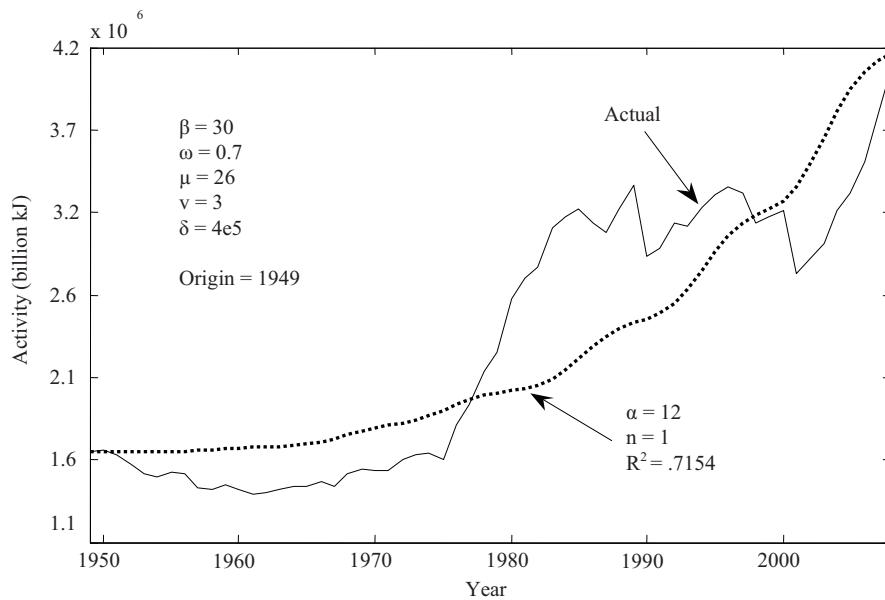


Figure A4.7. EIA U.S. Biomass Energy Production. U.S. biomass energy production (activity) scaled in billion kJ with actual and best-fit curves and common pattern equation parameters.

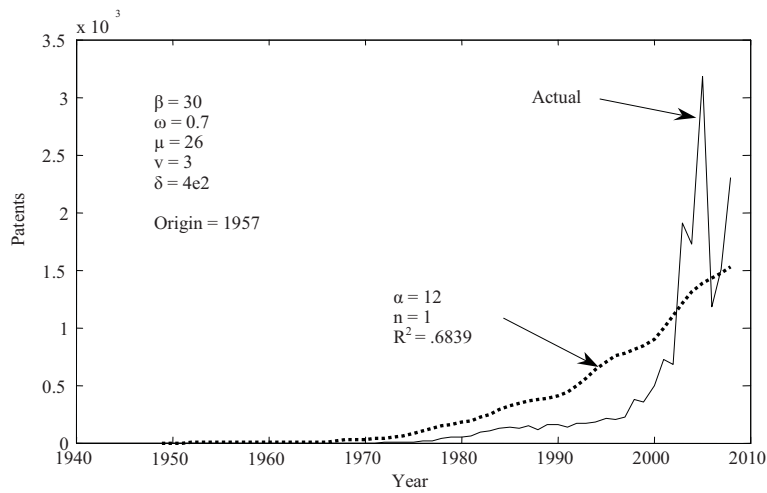


Figure A4.8. EPO Worldwide Patent Search: Biomass in title or abstract by date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

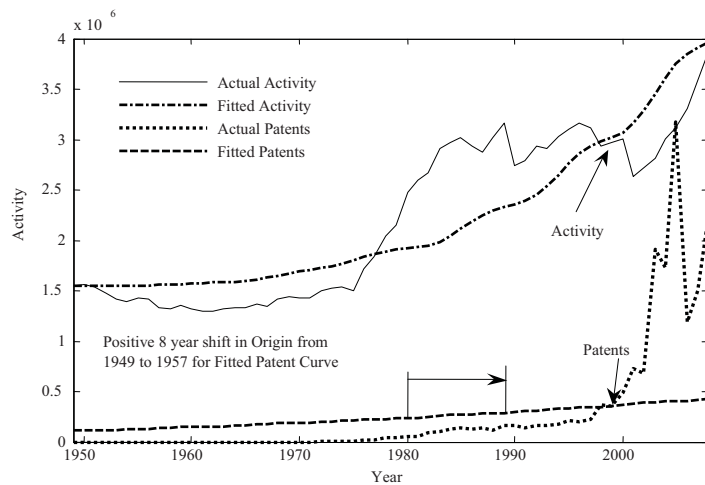


Figure A4.9. U.S. Biomass Energy Best-Fit Activity and Patents. Illustrates best-fit origin shift.

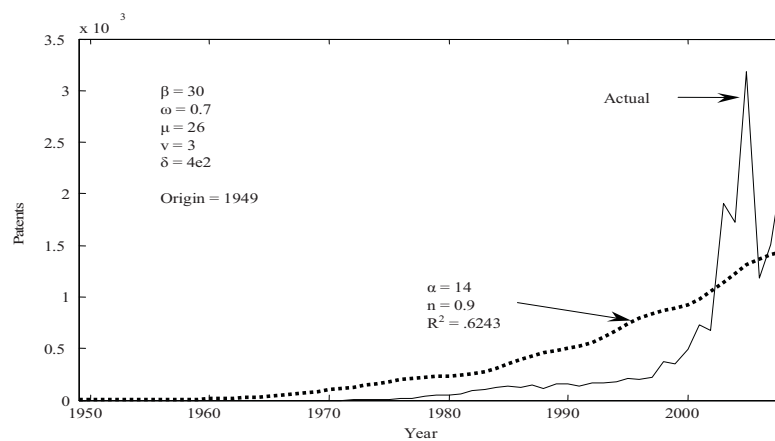


Figure A4.10. U.S. Biomass Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.5 U.S. Coal Energy Activity¹²¹ and Patents¹²²

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (pat.)
1900			1927			1954	11122283	267	1981	19387496	2289
1901			1928			1955	13049936	279	1982	19663906	2539
1902			1929			1956	14038182	312	1983	18195238	2675
1903			1930			1957	13779770	301	1984	20803773	2442
1904			1931			1958	11376158	232	1985	20388050	2266
1905			1932			1959	11370773	217	1986	20582487	2209
1906			1933			1960	11412355	285	1987	21248865	2200
1907			1934			1961	11021681	288	1988	21878209	1981
1908			1935			1962	11500091	262	1989	22534988	2022
1909			1936			1963	12500937	262	1990	23724363	1676
1910			1937			1964	13212694	270	1991	22826427	1842
1911			1938			1965	13773326	318	1992	22887309	1857
1912			1939			1966	14208401	281	1993	21454115	1556
1913			1940			1967	14585889	343	1994	23423198	1613
1914			1941			1968	14357155	327	1995	23346675	1561
1915			1942			1969	14625830	266	1996	24043606	1510
1916			1943			1970	15410453	297	1997	24591643	1596
1917			1944			1971	13910727	333	1998	25367685	1722
1918			1945			1972	14866728	373	1999	24576314	1781
1919			1946			1973	14761693	340	2000	23985929	1676
1920			1947			1974	14848546	356	2001	24842169	1660
1921			1948			1975	15813727	435	2002	23982510	1733
1922			1949	12632446	214	1976	16514640	648	2003	23308803	1843
1923			1950	14833442	162	1977	16621252	921	2004	24108964	2012
1924			1951	15212388	195	1978	15729848	1262	2005	24460374	2144
1925			1952	13434700	301	1979	18504260	1378	2006	25097933	3000
1926			1953	12953022	221	1980	19620601	1725	2007	24784843	4032
									2008	25167991	4778

Table A4.6 Correlation Eq.(A1.1) terms calculated from Table A4.5 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
1092767796	73886	2.13E+16	1.54E+08	1.59E+12	1.382E+15	63499265	2.44E+11	0.823513	67.81735

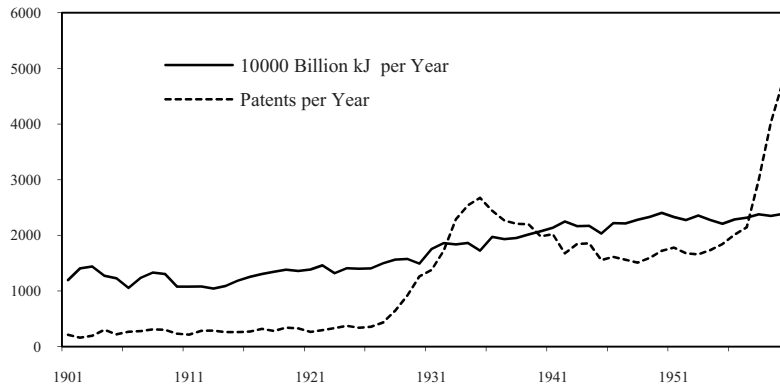


Figure A4.11. U.S. Coal Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

¹²¹ Activity represents United States production of coal energy, defined at eia.doe.gov as primary energy production from coal and in 1989 waste coal and 2001 refuse recovery. Data is in billions of Btu's as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ..

¹²² Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Coal was used as the keyword found in the patent title or abstract by year of publication.

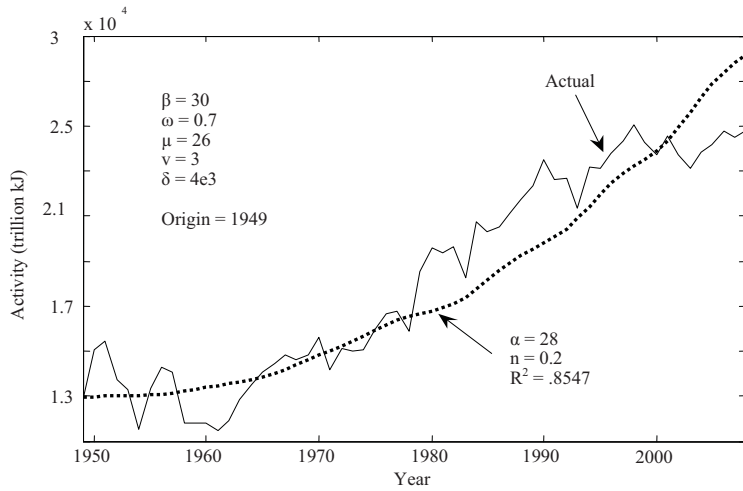


Figure A4.12. EIA U.S. Coal Energy Production. U.S. coal production (activity) scaled in trillions of kJ's with actual and best-fit curves and common pattern equation parameters.

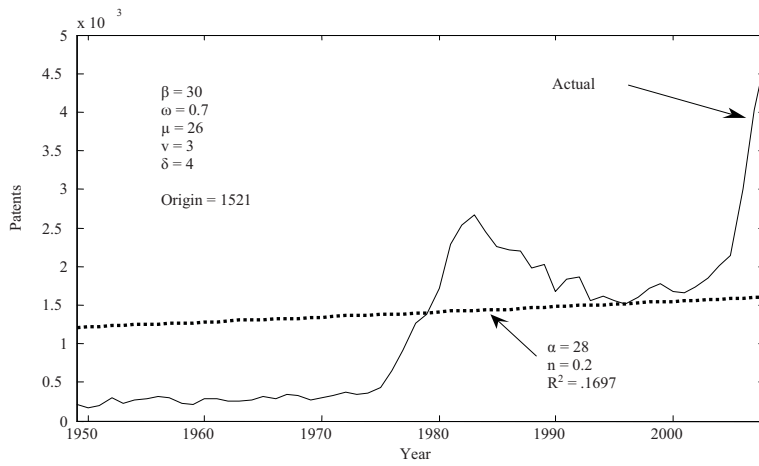


Figure A4.13. EPO Worldwide Patent Search: Coal in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

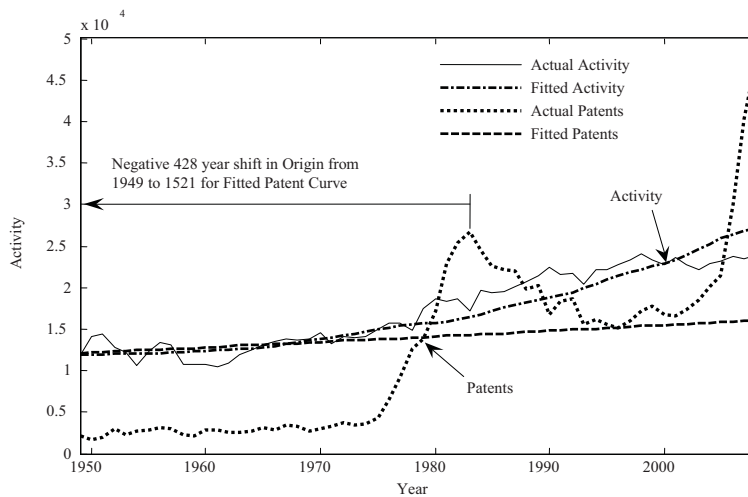


Figure A4.14. U.S. Coal Energy Best-Fit Activity and Patents. Illustrates best-fit origin shift.

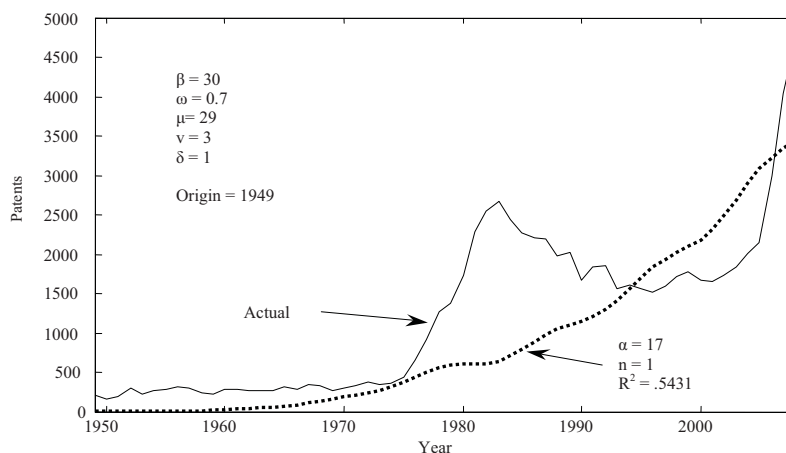


Figure A4.15. U.S. Coal Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.7 U.S. Fossil Fuel Energy Activity¹²³ and Patents¹²⁴

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (pat)
1900			1927			1954	35621368	820	1981	61748438	3838
1901			1928			1955	39418682	847	1982	60617996	4154
1902			1929			1956	41958882	1026	1983	57408839	4460
1903			1930			1957	42340826	1160	1984	62085860	4236
1904			1931			1958	39263220	1075	1985	60703354	4115
1905			1932			1959	41192703	1005	1986	59686869	4114
1906			1933			1960	42061918	1312	1987	60310921	4077
1907			1934			1961	42524028	1222	1988	61058103	3810
1908			1935			1962	44027139	1114	1989	60644226	4078
1909			1936			1963	46459226	1072	1990	61780380	3869
1910			1937			1964	48307343	1088	1991	61054672	3829
1911			1938			1965	49832822	1252	1992	60826087	4207
1912			1939			1966	52787312	1129	1993	58892299	3596
1913			1940			1967	55489974	1321	1994	61236037	3769
1914			1941			1968	57293027	1269	1995	60704840	3782
1915			1942			1969	59381275	1105	1996	61598522	3599
1916			1943			1970	62441305	1193	1997	62093750	3763
1917			1944			1971	61233846	1240	1998	62576342	4590
1918			1945			1972	62179489	1528	1999	60783280	4706
1919			1946			1973	61444773	1318	2000	60521144	5067
1920			1947			1974	59428950	1267	2001	61761121	4458
1921			1948			1975	57743603	1486	2002	60022709	5321
1922			1949	30329326	572	1976	57732655	1786	2003	59245148	5552
1923			1950	34353614	454	1977	58131325	2188	2004	58989213	6039
1924			1951	37760719	596	1978	58103194	2464	2005	58083775	6038
1925			1952	36900452	860	1979	61195917	2410	2006	59046401	7237
1926			1953	37293549	673	1980	62253306	3210	2007	59339357	9154
									2008	61126974	10376

¹²³ Activity represents United States production of fossil fuel energy, defined at eia.doe.gov as "primary energy from coal, natural gas (dry), crude oil, and natural gas plant liquids." Data is in billion Btu's as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ.

¹²⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Coal, petroleum and natural gas were used as keywords found in the patent title or abstract by year of publication.

Table A4.8 Correlation Eq.(A1.1) terms calculated from Table A4.7 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
3270432426	176896	1.83E+17	7.95E+08	1.04E+13	5.185E+15	2.74E+08	7.1E+11	0.595587	35.47237

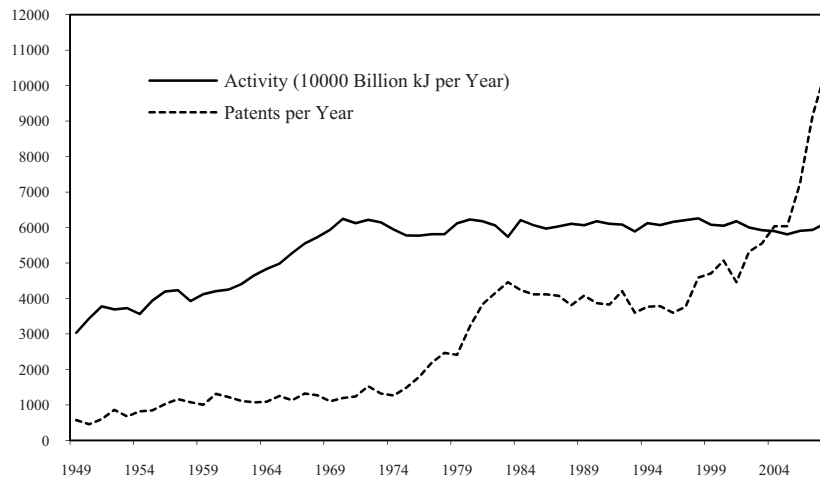


Figure A4.16. U.S. Fossil Fuel Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

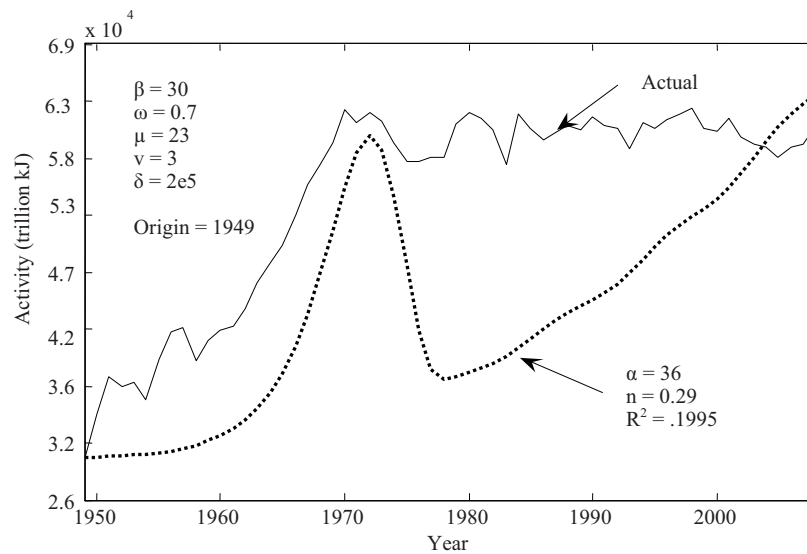


Figure A4.17. EIA U.S. Fossil Fuel Energy Production. U.S. fossil fuel power production (activity) scaled in trillion kJ with actual and best-fit curves and common pattern equation parameters.

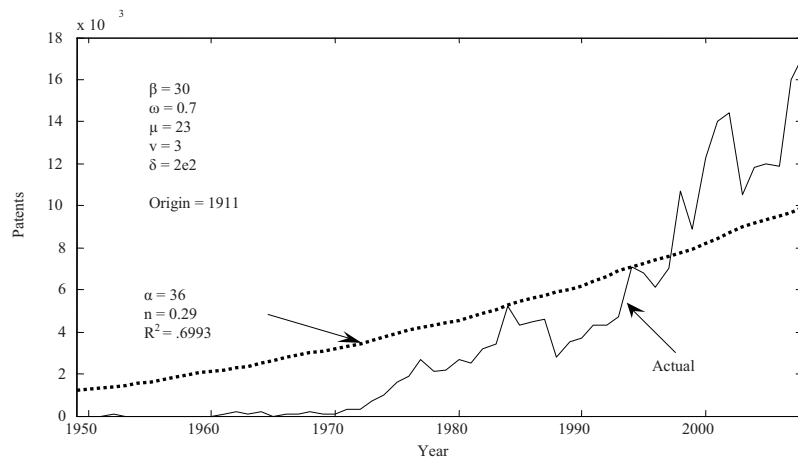


Figure A4.18. EPO Worldwide Patent Search: Fossil Fuel or Natural Gas or Coal in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

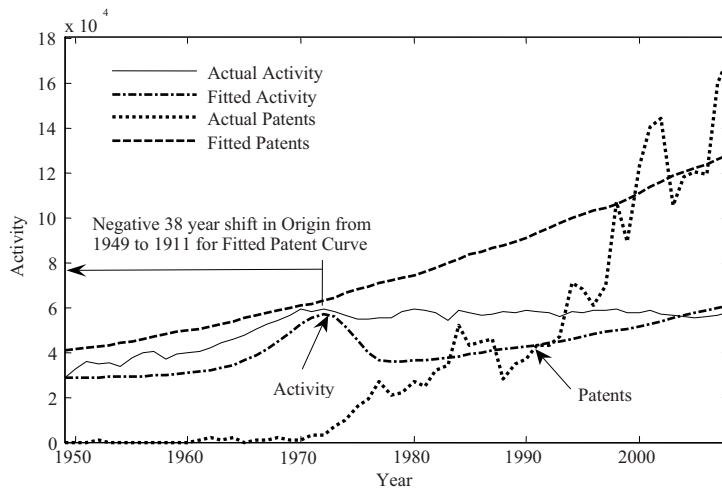


Figure A4.19. U.S. Fossil Fuel Power Best-fit Activity and Patents. Illustrates best-fit origin shift.

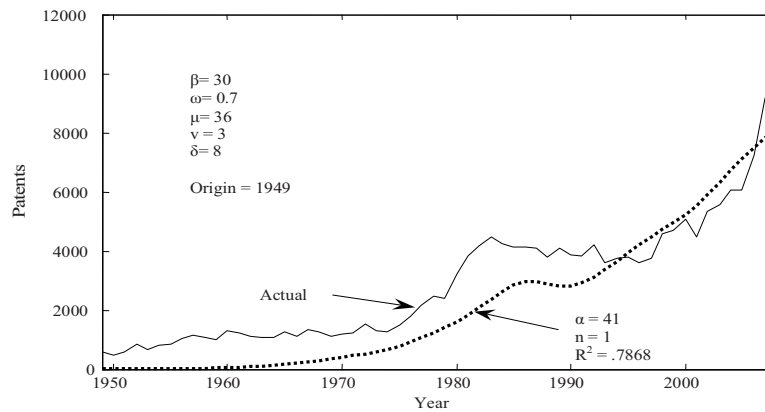


Figure A4.20. U.S. Fossil Fuel Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.9 U.S. Geothermal Energy Activity¹²⁵ and Patents¹²⁶

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981	129810.4	30
1901			1928			1955			1982	110507	52
1902			1929			1956			1983	136452.6	46
1903			1930			1957			1984	173965.3	41
1904			1931			1958			1985	209187.5	48
1905			1932			1959			1986	231232.8	38
1906			1933			1960	816.57	0	1987	241720.5	34
1907			1934			1961	2300.955	0	1988	229241	31
1908			1935			1962	2459.205	0	1989	334607	27
1909			1936			1963	3930.93	0	1990	354270.1	27
1910			1937			1964	4768.6	1	1991	365290.6	24
1911			1938			1965	4427.835	0	1992	368521	22
1912			1939			1966	4399.35	0	1993	383720.4	13
1913			1940			1967	7264.73	1	1994	356703.9	20
1914			1941			1968	9933.88	0	1995	310057.1	26
1915			1942			1969	14011.46	0	1996	332883.1	36
1916			1943			1970	11971.09	0	1997	342831.7	45
1917			1944			1971	12514.41	1	1998	346359.7	36
1918			1945			1972	33210.35	4	1999	349119.5	37
1919			1946			1973	44948.28	3	2000	334219.8	40
1920			1947			1974	56081.69	11	2001	328383.5	53
1921			1948			1975	74011.42	13	2002	346364.9	54
1922			1949			1976	82452.47	29	2003	348734.5	56
1923			1950			1977	81675.99	28	2004	359841.5	66
1924			1951			1978	67889.25	32	2005	361417.7	54
1925			1952			1979	88396.34	31	2006	361734.2	85
1926			1953			1980	115813.7	35	2007	367910.2	116
									2008	378214.3	139

Table A4.10. Correlation Eq.(A1.1) terms calculated from Table A4.9 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
9216580.2	1485	2.82E+12	85903	4.19E+08	1.082E+12	40898.41	1.4E+08	0.665995	44.35495

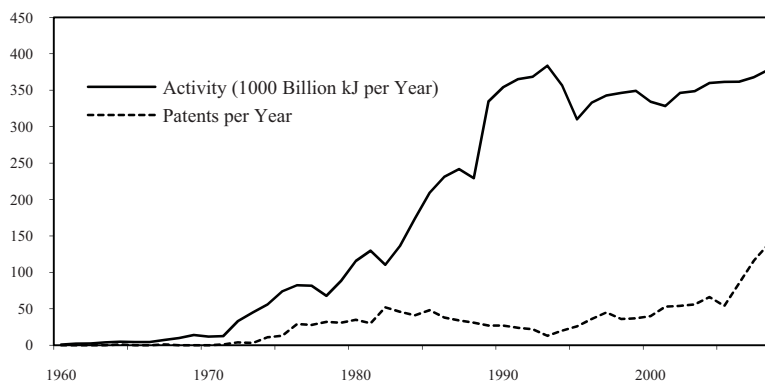


Figure A4.21. U.S. Geothermal Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

Activity represents U.S. production of geothermal energy, defined at usgs.gov as “...electricity generation (converted to Btu using the geothermal energy plants heat rate), and geothermal heat pump and direct energy use energy.” Data is in billion Btu’s as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ.

¹²⁶ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Geothermal and (power or energy) were used as keywords found in the patent title or abstract by year of publication.

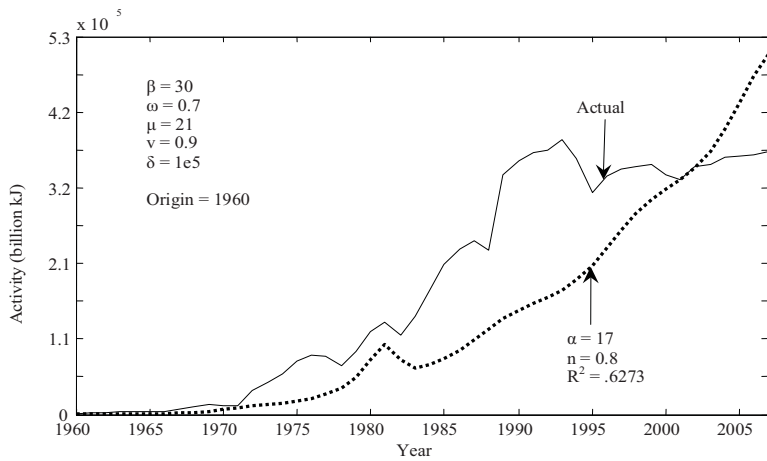


Figure A4.22. EIA U.S. Geothermal Energy Production. U.S. geothermal energy production (activity) scaled in billion kJ with actual and best-fit curves and common pattern equation parameters.

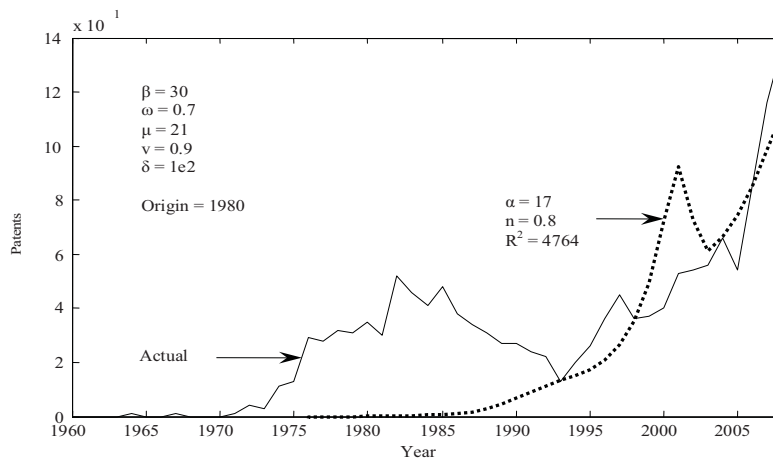


Figure A4.23. EPO Worldwide Patent Search: Geothermal and (Energy or Power) in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

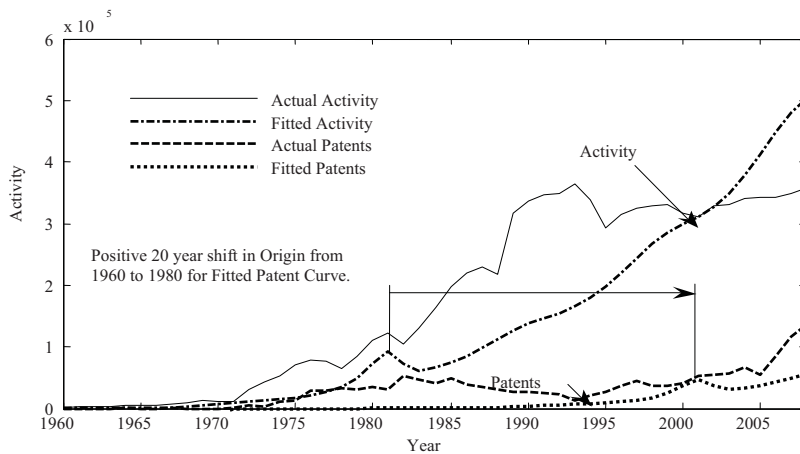


Figure A4.24. U.S. Geothermal Energy Best-Fit Activity and Patents. Illustrates best-fit origin shift.

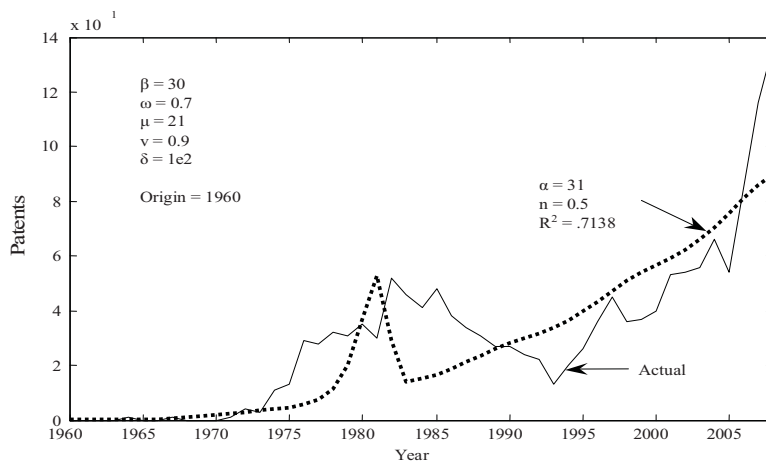


Figure A4.25. U.S. Geothermal Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.11 U.S. Hydroelectric Energy Activity¹²⁷ and Patents¹²⁸

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	1366571	0	1981	2771758	36
1901			1928			1955	1366643	2	1982	3281886	55
1902			1929			1956	1441885	3	1983	3544896	25
1903			1930			1957	1523191	1	1984	3402740	32
1904			1931			1958	1599927	2	1985	2985043	35
1905			1932			1959	1556207	1	1986	3086535	24
1906			1933			1960	1616015	1	1987	2647681	39
1907			1934			1961	1664745	1	1988	2345936	40
1908			1935			1962	1825222	0	1989	2851449	30
1909			1936			1963	1780212	0	1990	3061623	43
1910			1937			1964	1895746	1	1991	3031023	40
1911			1938			1965	2069372	6	1992	2630523	55
1912			1939			1966	2071827	4	1993	2906071	44
1913			1940			1967	2358397	0	1994	2696874	64
1914			1941			1968	2360372	2	1995	3221334	41
1915			1942			1969	2661223	2	1996	3607604	46
1916			1943			1970	2646715	6	1997	3658660	65
1917			1944			1971	2838272	4	1998	3313539	40
1918			1945			1972	2878184	2	1999	3283913	69
1919			1946			1973	2875755	4	2000	2825172	66
1920			1947			1974	3192463	3	2001	2253067	68
1921			1948			1975	3170380	1	2002	2702462	79
1922			1949	1431846	0	1976	2991146	2	2003	2838656	105
1923			1950	1422488	2	1977	2344918	6	2004	2703528	90
1924			1951	1430914	2	1978	2951668	12	2005	2716457	105
1925			1952	1473141	1	1979	2945339	15	2006	2883380	113
1926			1953	1419923	0	1980	2914645	20	2007	2458621	157
									2008	2464333	206

¹²⁷ Activity represents United States production of hydroelectric energy, defined at eia.doe.gov as “Conventional hydroelectricity net generation...” Data is in billion Btu’s as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ.

¹²⁸ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Hydroelectric was used as the keyword found in the patent title or abstract by year of publication.

Table A4.12 Correlation Eq.(A1.1) terms calculated from Table A4.11 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
150260117	1918	4.03E+14	164888	5.4E+09	2.653E+13	103575.9	6E+08	0.361657	13.07958

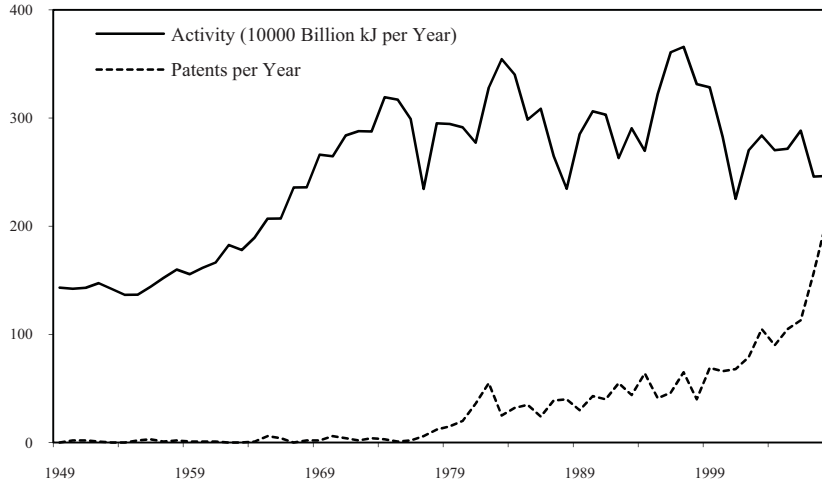


Figure A4.26. U.S. Hydroelectric Power Activity and Patent. Data illustrates correlation. Activity scaled to fit plot.

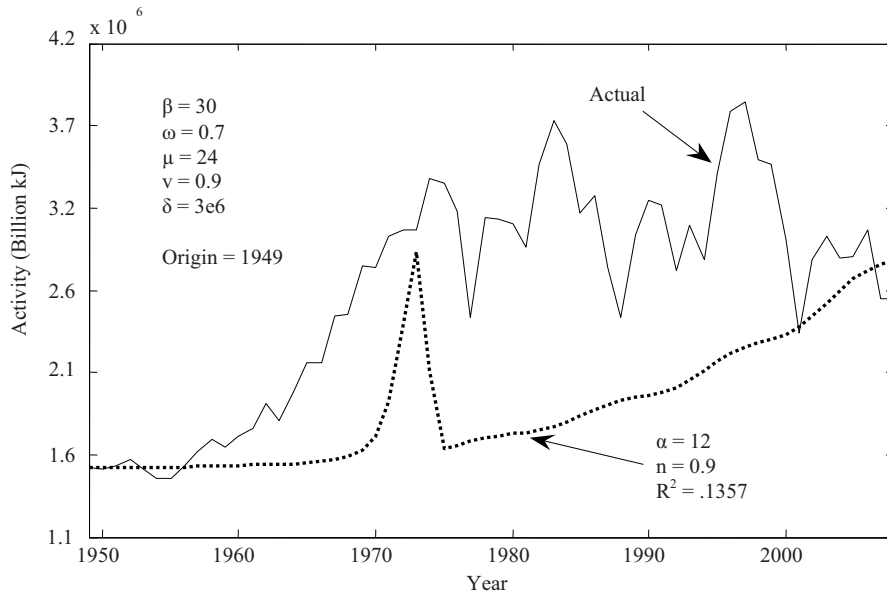


Figure A4.27. EIA U.S. Hydroelectric Energy Production. U.S. hydroelectric energy (activity) production scaled in billion kJ with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data or origin shift was obtained suggesting Stage IV.

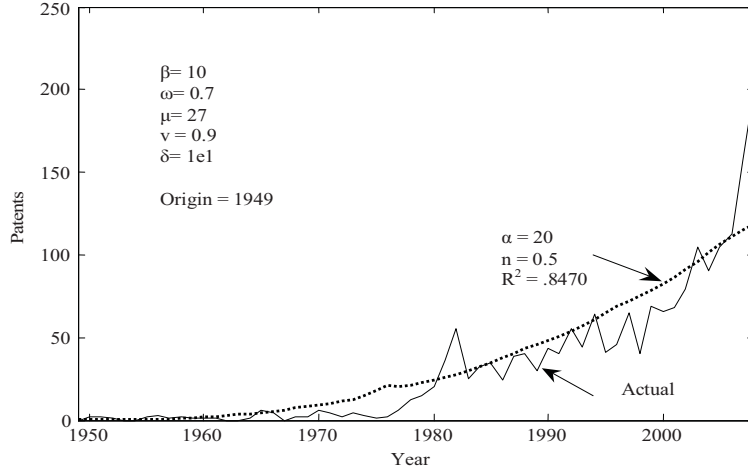


Figure A4.28. U.S. Hydroelectric Power Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.13 U.S. Natural Gas Energy Activity¹²⁹ and Patents¹³⁰

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (pat.)
1900			1927			1954	10333674	58	1981	23216892	284
1901			1928			1955	11166530	56	1982	21638152	309
1902			1929			1956	11905545	42	1983	19810258	310
1903			1930			1957	12548239	63	1984	21397197	334
1904			1931			1958	12901710	57	1985	20278314	373
1905			1932			1959	14068012	76	1986	19717848	328
1906			1933			1960	14893548	111	1987	20415151	352
1907			1934			1961	15459810	89	1988	20950640	361
1908			1935			1962	16151448	74	1989	21105926	422
1909			1936			1963	17114378	90	1990	21628417	473
1910			1937			1964	18042085	93	1991	21663872	462
1911			1938			1965	18629439	115	1992	21878654	560
1912			1939			1966	20052029	91	1993	22146601	525
1913			1940			1967	21226961	140	1994	22934638	559
1914			1941			1968	22565464	152	1995	22707639	637
1915			1942			1969	24123701	98	1996	23077258	627
1916			1943			1970	25507577	148	1997	23092897	656
1917			1944			1971	26188618	141	1998	23245595	884
1918			1945			1972	26169364	209	1999	23071112	952
1919			1946			1973	26117715	200	2000	23497418	1144
1920			1947			1974	24983786	164	2001	23961903	1118
1921			1948			1975	23225445	191	2002	23207896	1206
1922			1949	6426049	43	1976	23006770	181	2003	23248886	1269
1923			1950	7443872	21	1977	23096381	245	2004	22744423	1369
1924			1951	8794588	30	1978	22925843	217	2005	22057842	1340
1925			1952	9454245	58	1979	23592362	203	2006	22553307	1347
1926			1953	9917543	48	1980	23380098	309	2007	23243847	1699
									2008	24860924	1827

¹²⁹ Activity represents United States production of natural gas energy, defined at eia.doe.gov as primary energy produced from dry natural gas and natural gas plant liquids. Data is in billions of Btu's as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ..

¹³⁰ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Natural and gas or methane or ethane were used as keywords found in the patent title or abstract by year of publication.

Table A4.14 Correlation Eq.(A1.1) terms calculated from Table A4.13 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
1.195E+09	25540	2.54E+16	23262650	5.75E+11	1.577E+15	12391123	6.61E+10	0.472812	22.35514

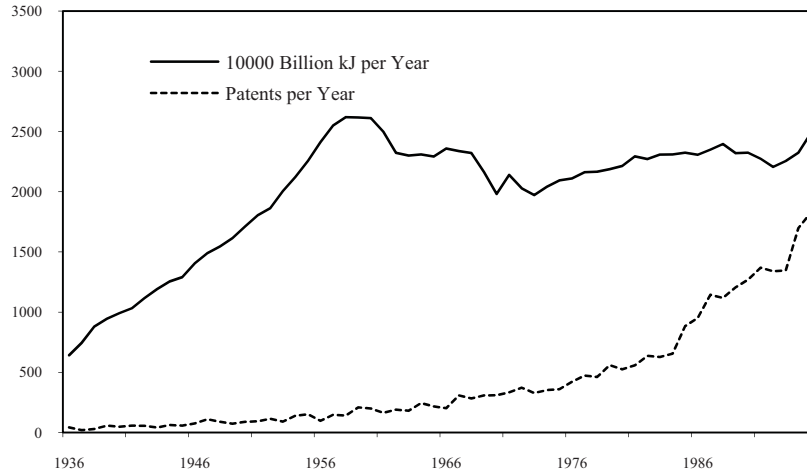


Figure A4.29. U.S. Natural Gas Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

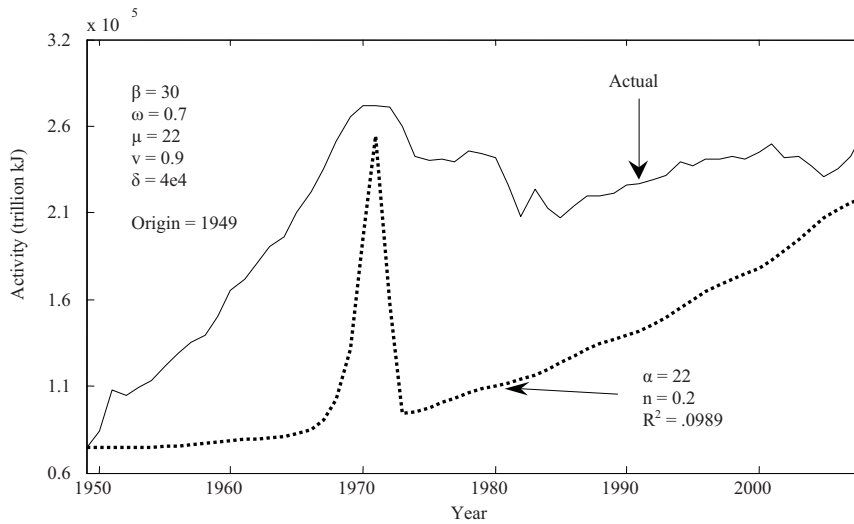


Figure A4.30. EIA U.S. Natural Gas Energy Production. U.S. natural gas energy production (activity) scaled in billion kJ with actual and best-fit curves and common pattern equation parameters.

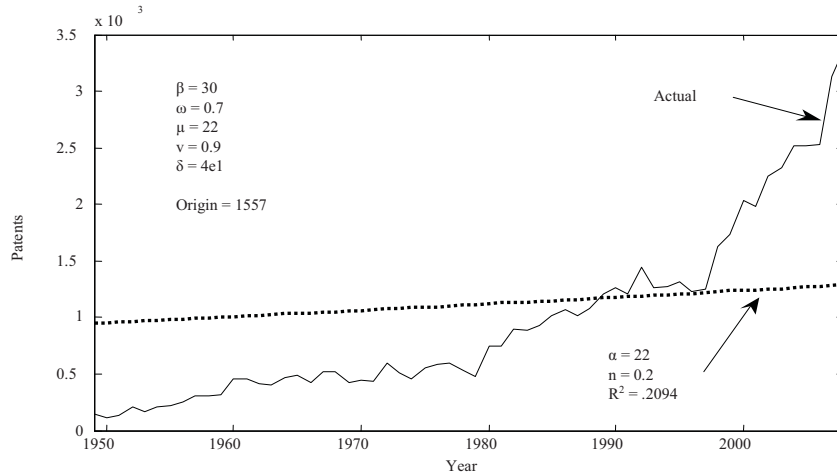


Figure A4.31. EPO Worldwide Patent Search: Natural Gas or Methane or Ethane in Title or Abstract by Date of Production. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

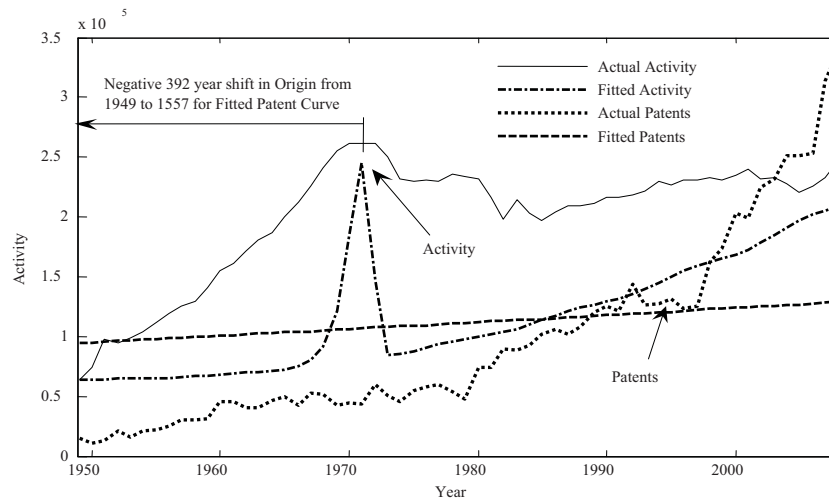


Figure A4.32. U.S. Natural Gas Energy Best-Fit Activity and Patents. Illustrates best-fit origin shift.

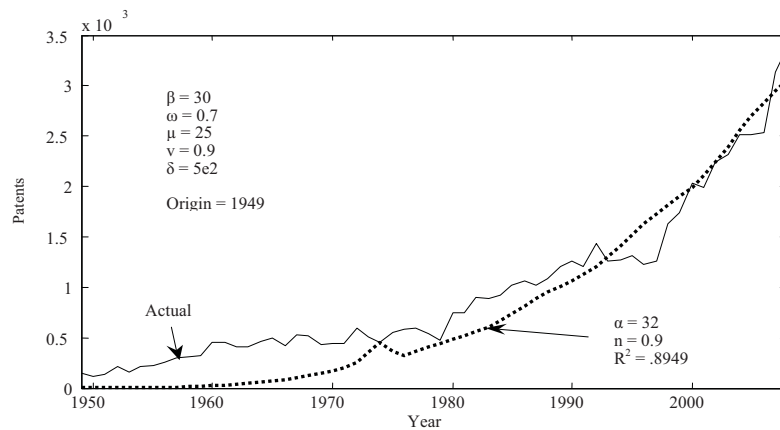


Figure A4.33. U.S. Natural Gas Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.15 U.S. Nuclear Energy Activity¹³¹ and Patents¹³²

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981	3173006	1995
1901			1928			1955			1982	3303361	1957
1902			1929			1956			1983	3378689	1684
1903			1930			1957	118.16	246	1984	3747920	1745
1904			1931			1958	2020.325	402	1985	4299719	2424
1905			1932			1959	2307.285	490	1986	4621015	2572
1906			1933			1960	6357.43	852	1987	5015399	2715
1907			1934			1961	20760.29	844	1988	5894251	2202
1908			1935			1962	27845.67	855	1989	5910280	2300
1909			1936			1963	40245.09	911	1990	6440089	2106
1910			1937			1964	42009.05	886	1991	6775349	1983
1911			1938			1965	45538.02	871	1992	6835562	1993
1912			1939			1966	67686.69	871	1993	6763076	1781
1913			1940			1967	93321.08	907	1994	7062040	1826
1914			1941			1968	149318.4	941	1995	7464585	1619
1915			1942			1969	162176.7	758	1996	7476441	1632
1916			1943			1970	252511.1	766	1997	6959827	1729
1917			1944			1971	435650.6	821	1998	7456538	1896
1918			1945			1972	615858.4	828	1999	8028820	1912
1919			1946			1973	960236.7	752	2000	8294778	1959
1920			1947			1974	1342048	718	2001	8474495	1977
1921			1948			1975	2004287	693	2002	8590959	2245
1922			1949			1976	2227233	877	2003	8396595	2327
1923			1950			1977	2850359	1169	2004	8674194	2307
1924			1951			1978	3190453	1190	2005	8608830	2242
1925			1952			1979	2928497	1337	2006	8665600	2130
1926			1953			1980	2889823	1558	2007	8922961	2297
									2008	8920274	2461

Table A4.16 Correlation Eq.(A1.1) terms calculated from Table A4.15 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
208511317	78559	1.41E+15	1.42E+08	4.13E+11	5.71E+14	23119498	9.81E+10	0.854099	72.94843

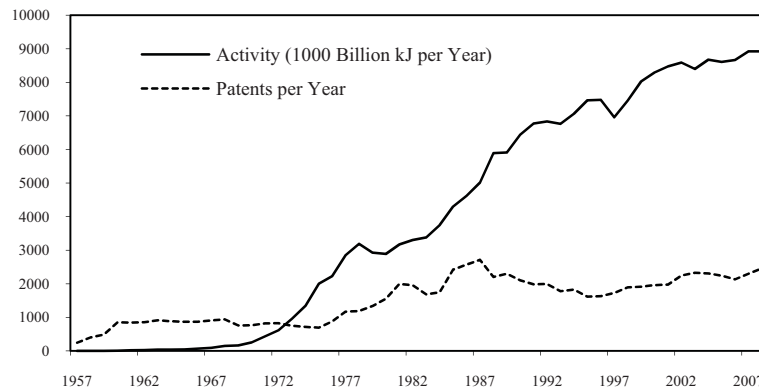


Figure A4.34. U.S. Nuclear Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

¹³¹ Activity represents United States production of nuclear electric power. Data is in billion Btu's as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ.

¹³² Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Nuclear or uranium were used as keywords found in the patent title or abstract by year of publication.

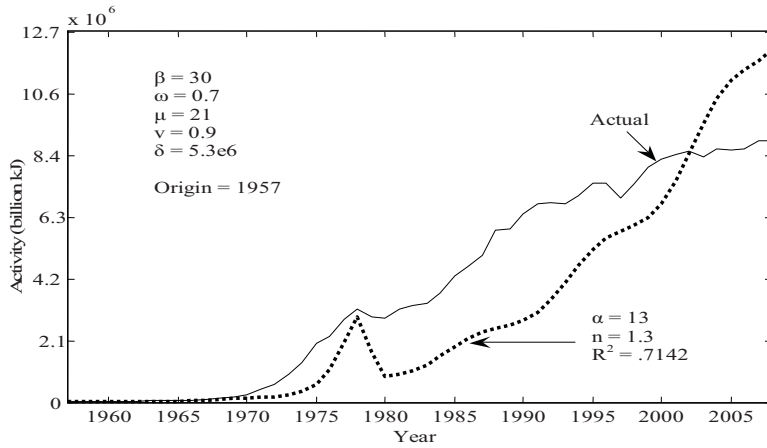


Figure A4.35. EIA U.S. Nuclear Energy Production. U.S. nuclear energy production (activity) scaled in billion kJ with actual and best-fit curves and common pattern equation parameters.

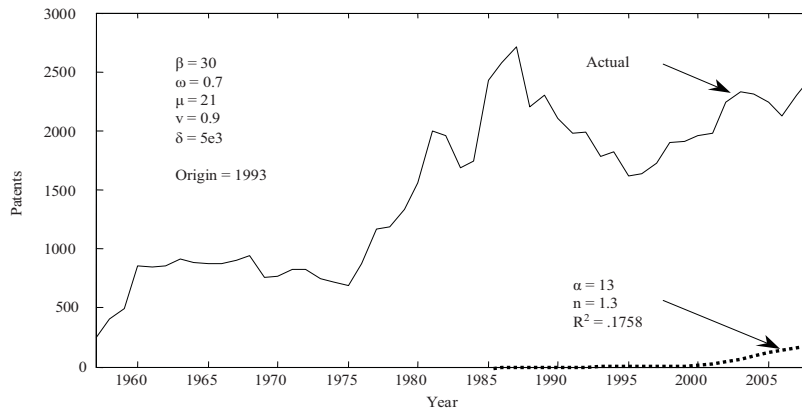


Figure A4.36. EPO Worldwide Patent Search: Nuclear or Uranium in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

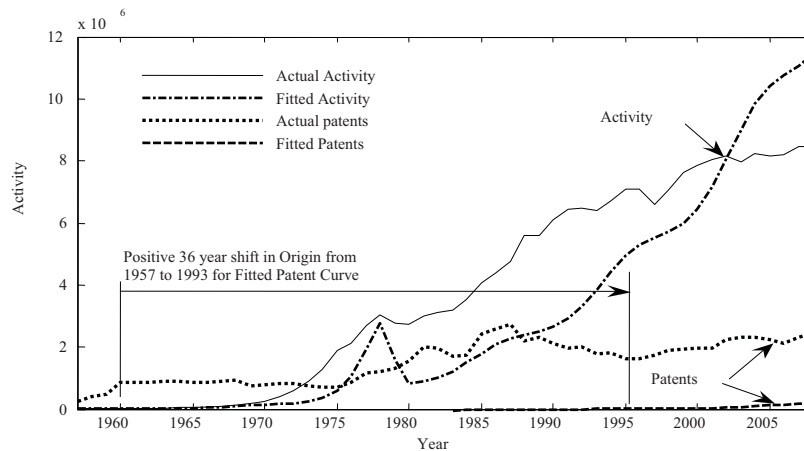


Figure A4.37. U.S. Nuclear Energy Best-Fit Activity and Patents. Illustrates best-fit origin shift.

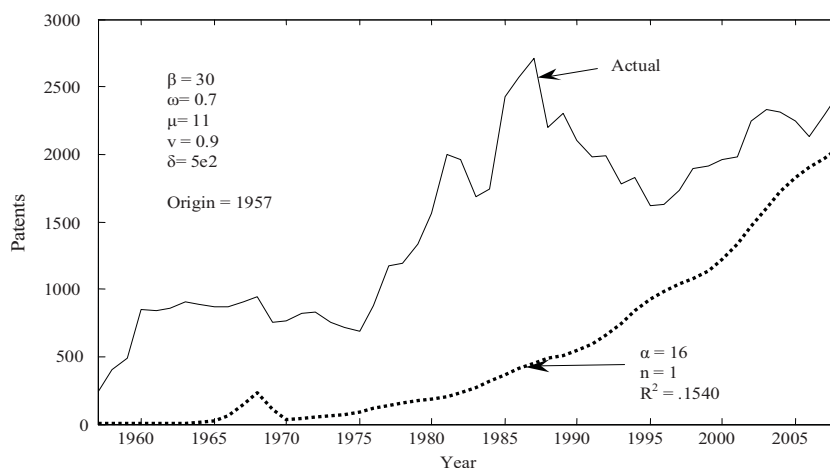


Figure A4.38. U.S. Nuclear Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.17 U.S. Oil Energy Activity¹³³ and Patents¹³⁴

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (pat.)
1900			1927			1954	14165411	343	1981	19144050	807
1901			1928			1955	15202215	345	1982	19315939	873
1902			1929			1956	16015154	460	1983	19403343	895
1903			1930			1957	16012817	558	1984	19884890	870
1904			1931			1958	14985352	537	1985	20036989	834
1905			1932			1959	15753916	471	1986	19386534	841
1906			1933			1960	15756015	575	1987	18646906	860
1907			1934			1961	16042537	482	1988	18229253	749
1908			1935			1962	16375600	400	1989	17003312	852
1909			1936			1963	16843912	405	1990	16427600	936
1910			1937			1964	17052564	355	1991	16564371	978
1911			1938			1965	17430057	441	1992	16060124	920
1912			1939			1966	18526881	424	1993	15291583	787
1913			1940			1967	19677126	454	1994	14878202	883
1914			1941			1968	20370408	426	1995	14650525	911
1915			1942			1969	20631745	411	1996	14477658	859
1916			1943			1970	21523277	450	1997	14409211	913
1917			1944			1971	21134500	470	1998	13963062	1241
1918			1945			1972	21143396	561	1999	13135854	1189
1919			1946			1973	20565365	470	2000	13037797	1357
1920			1947			1974	19596618	455	2001	12957050	1312
1921			1948			1975	18704430	503	2002	12832302	1342
1922			1949	11270831	214	1976	18211245	557	2003	12687458	1367
1923			1950	12076299	180	1977	18413693	673	2004	12135825	1514
1924			1951	13753744	270	1978	19447504	667	2005	11565559	1379
1925			1952	14011507	346	1979	19099296	609	2006	11395163	1704
1926			1953	14422985	289	1980	19252607	741	2007	11310668	1989
									2008	11098059	2190

¹³³ Activity represents United States production of oil energy, defined at eia.doe.gov as primary energy production of crude oil and lease condensate. Data is in thousand barrels as reported by the Energy Information Administration (EIA) at eia.doe.gov.

¹³⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Petroleum was used as the keyword found in the patent title or abstract by year of publication.

Table A4.18 Correlation Eq.(A1.1) terms calculated from Table A4.17 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
983398293	45894	1.66E+16	46031010	7.15E+11	5.274E+14	10926689	-3.7E+10	-0.48622	23.64057

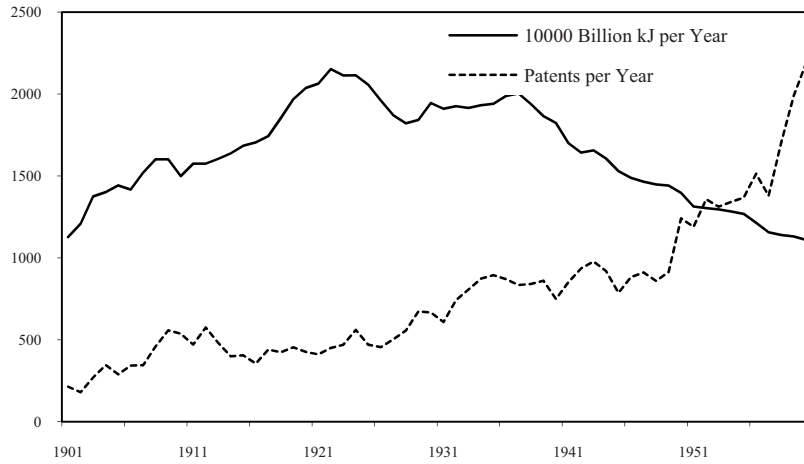


Figure A4.39. U.S. Oil Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

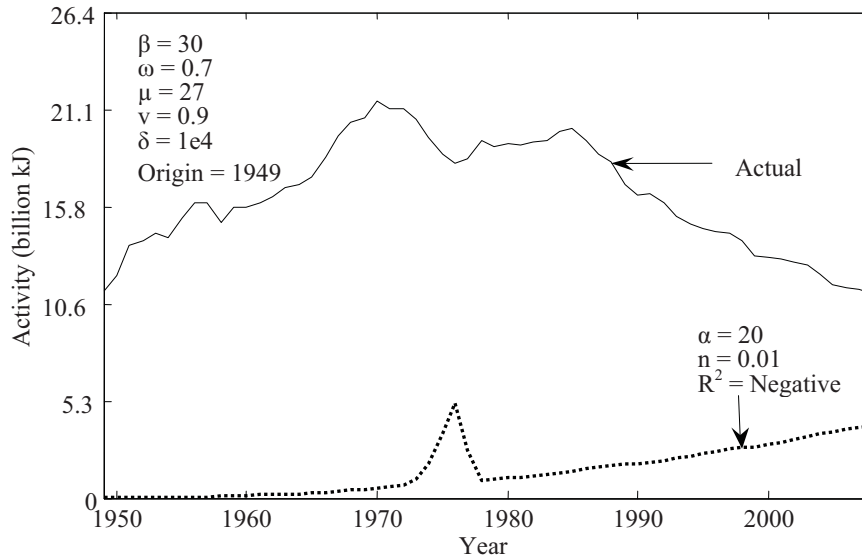


Figure A4.40. EIA U.S. Oil Energy Production. U.S. oil energy production (activity) scaled billion kJ with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

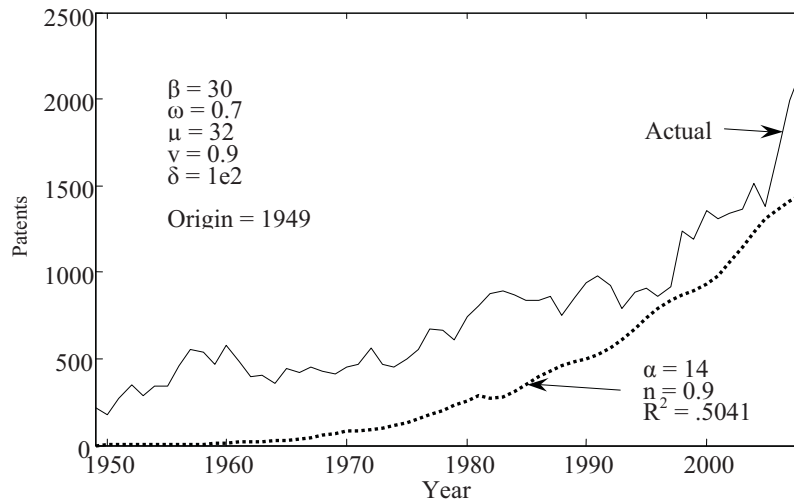


Figure A4.41. U.S. Oil Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.19 U.S. Renewable Energy Activity¹³⁵ and Patents¹³⁶

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (pat.)
1900			1927			1954	290.55744	0	1981	577.88195	1
1901			1928			1955	293.71063	0	1982	636.63542	4
1902			1929			1956	300.7364	0	1983	692.32582	1
1903			1930			1957	300.58997	0	1984	688.23199	2
1904			1931			1958	307.542	0	1985	652.70529	1
1905			1932			1959	306.09126	0	1986	656.71925	2
1906			1933			1960	308.9693	0	1987	605.69249	5
1907			1934			1961	311.58433	0	1988	587.66011	8
1908			1935			1962	329.02433	0	1989	674.53187	4
1909			1936			1963	326.88078	0	1990	654.9746	7
1910			1937			1964	340.5157	0	1991	658.32897	3
1911			1938			1965	358.4928	0	1992	632.48632	3
1912			1939			1966	362.35811	0	1993	660.92332	13
1913			1940			1967	389.69579	0	1994	649.5692	13
1914			1941			1968	398.53058	0	1995	707.35693	11
1915			1942			1969	432.73473	0	1996	756.18634	8
1916			1943			1970	430.00291	0	1997	757.58959	9
1917			1944			1971	450.30934	0	1998	702.53062	14
1918			1945			1972	464.03215	0	1999	705.01599	10
1919			1946			1973	467.69427	0	2000	660.61726	9
1920			1947			1974	503.17117	0	2001	560.99878	16
1921			1948			1975	498.32862	0	2002	622.31475	38
1922			1949	3137553	0	1976	503.00206	1	2003	648.62972	45
1923			1950	3141492	0	1977	448.26971	1	2004	659.16041	46
1924			1951	3121180	0	1978	531.60796	0	2005	676.24223	75
1925			1952	3101891	0	1979	545.05298	0	2006	723.40263	88
1926			1953	2987190	0	1980	578.71181	2	2007	717.40095	201
									2008	771.80751	313

¹³⁵ Activity represents United States production of renewable energy, defined at eia.doe.gov as "...hydroelectric, geothermal, solar, wind and biomass power." Data is in billion Btu's as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ.

¹³⁶ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Renewable and energy were used as the keywords found in the patent title or abstract by year of publication.

Table A4.20 Correlation Eq.(A1.1) terms calculated from Table A4.19 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
310250481	954	1.76E+15	158766	6.81E+09	1.51E+14	143597.4	1.88E+09	0.403048	16.24474

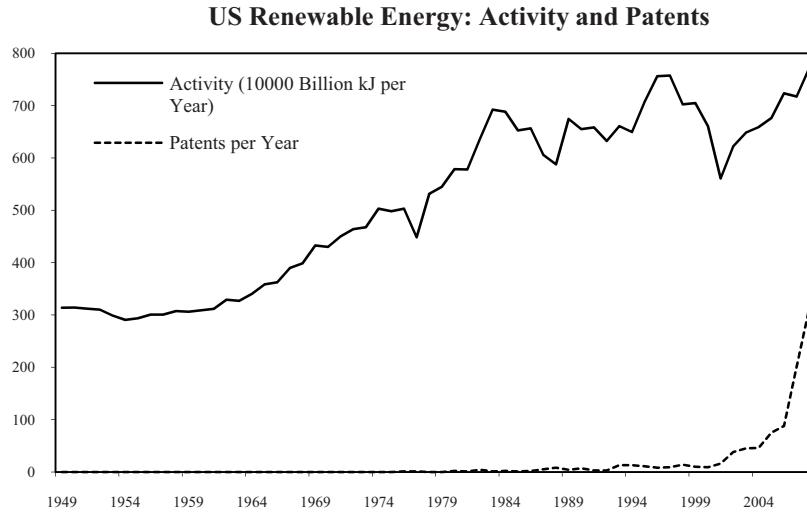


Figure A4.42. U.S. Renewable Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

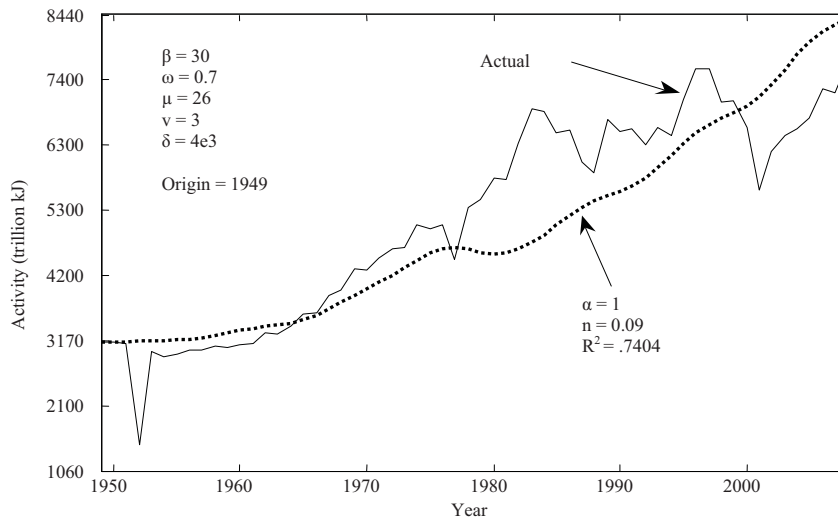


Figure A4.43. EIA U.S. Renewable Energy Production. U.S. renewable energy production (activity) scaled trillion kJ with actual and best-fit curves and common pattern equation parameters.

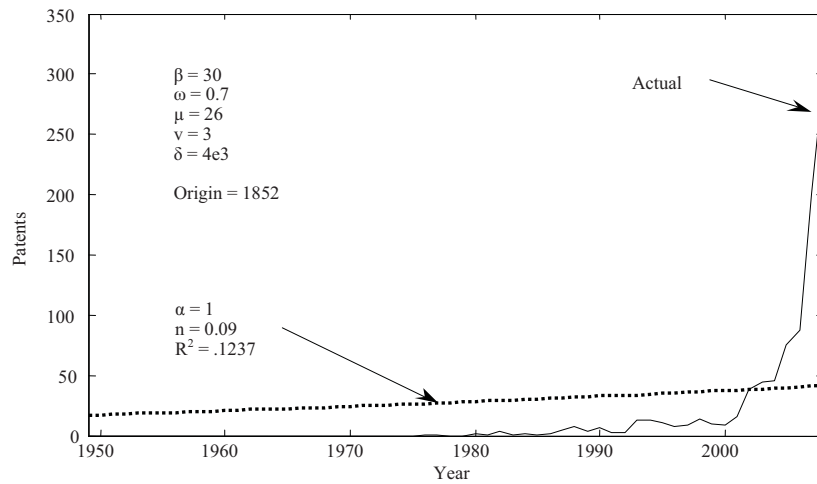


Figure A4.44. EPO Worldwide Patent Search: Renewable and Energy in Title or Abstract by date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

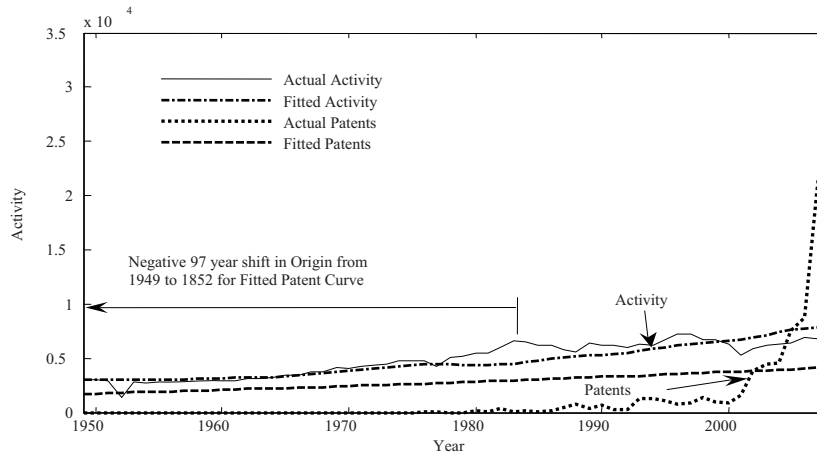


Figure A4.45. U.S. Renewable Energy Best-Fit Activity and Patents. Illustrates best-fit origin shift.

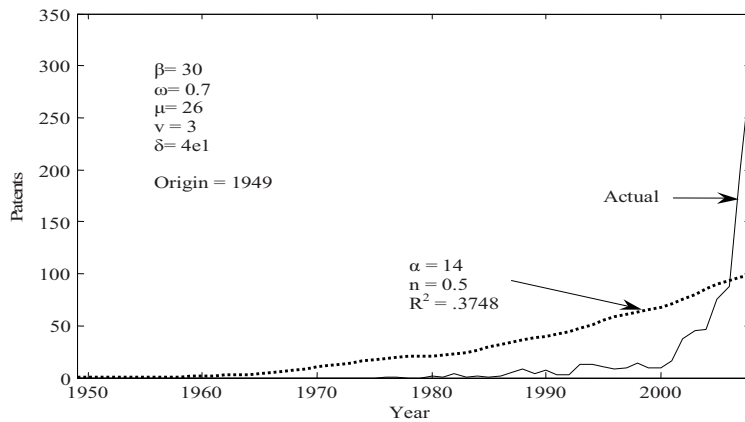


Figure A4.46. U.S. Renewable Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.21 U.S. Solar Energy Activity¹³⁷ and Patents¹³⁸

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981		
1901			1928			1955			1982		
1902			1929			1956			1983		
1903			1930			1957			1984	58.025	2323
1904			1931			1958			1985	117.105	2155
1905			1932			1959			1986	155.085	1928
1906			1933			1960			1987	114.995	1656
1907			1934			1961			1988	99.17	1415
1908			1935			1962			1989	58332.01	1507
1909			1936			1963			1990	63002.49	1457
1910			1937			1964			1991	66135.84	1445
1911			1938			1965			1992	67399.73	1740
1912			1939			1966			1993	70113.19	1718
1913			1940			1967			1994	72318.14	1939
1914			1941			1968			1995	73699.14	1990
1915			1942			1969			1996	74728.82	2043
1916			1943			1970			1997	74100.04	2186
1917			1944			1971			1998	73625.29	2597
1918			1945			1972			1999	72576.62	3025
1919			1946			1973			2000	70039.34	3534
1920			1947			1974			2001	69053.97	3761
1921			1948			1975			2002	67932.51	4105
1922			1949			1976			2003	67119.1	4252
1923			1950			1977			2004	68047.5	4378
1924			1951			1978			2005	69767.15	5016
1925			1952			1979			2006	76194.21	5605
1926			1953			1980			2007	85394.87	7327
									2008	96008.17	9165

Table A4.22. Correlation Eq.(A1.1) terms calculated from Table A4.21 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
1436132.5	78267	1.04E+11	3.39E+08	5.18E+09	2.174E+10	93491639	6.82E+08	0.478581	22.90399

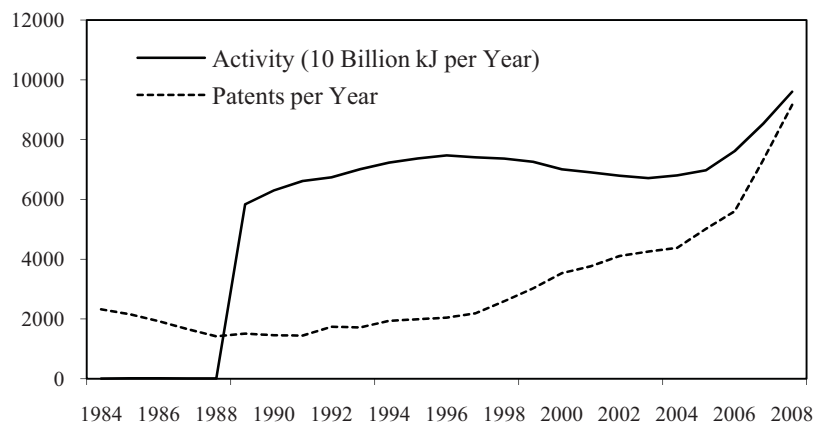


Figure A4.47. U.S. Solar Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

Activity represents U.S. production of solar energy, defined at usgs.gov as “...solar thermal and photovoltaic electricity net generation (converted to Btu using the fossil-fueled plants heat rate, and solar thermal direct use energy.” Data is in billion Btu’s as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ.

¹³⁸ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Solar was used as the keyword found in the patent title or abstract by year of publication.

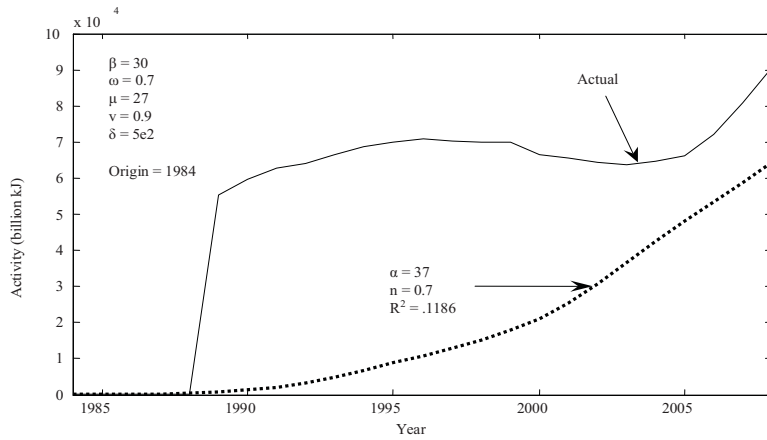


Figure A4.48. EIA U.S. Solar Energy Production. U.S. solar energy (activity) production scaled in billion kJ with actual and best-fit curves and common pattern equation parameters.

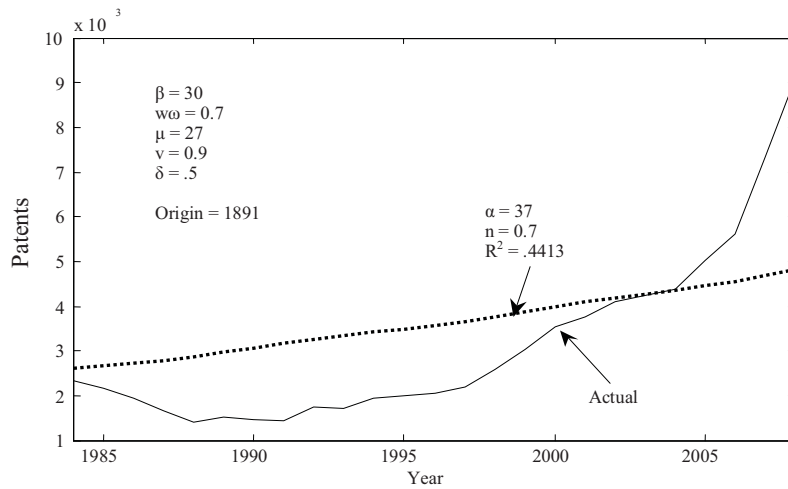


Figure A4.49. EPO Worldwide Patent Search: Solar in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

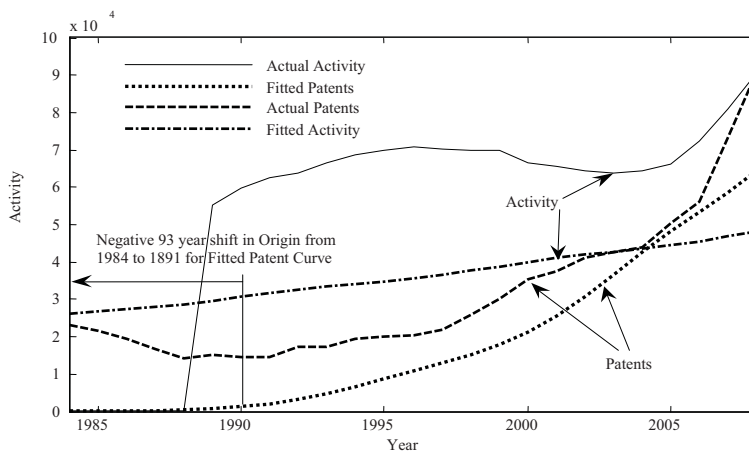


Figure A4.50. U.S. Solar Energy Best-Fit Activity and Patents. Illustrates best-fit origin shift.

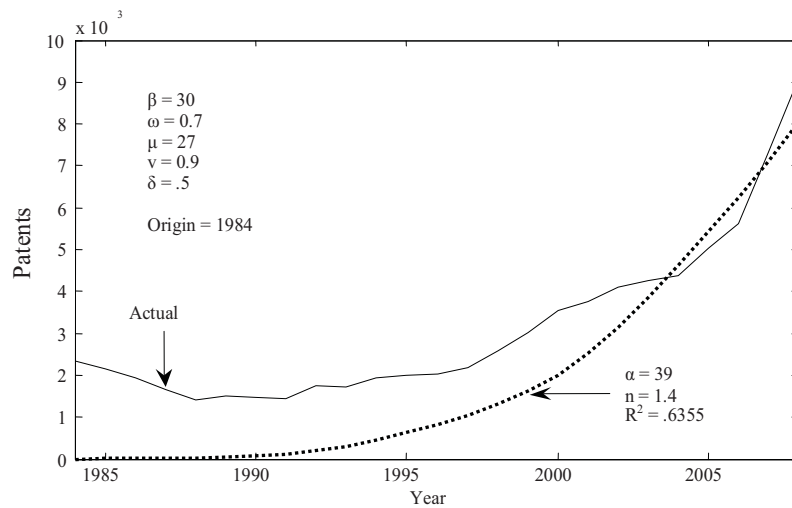


Figure A4.51. U.S. Solar Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.23 U.S. Total Energy Activity¹³⁹ and Patents¹⁴⁰

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (pat)
1900			1927			1954	38526944	820	1981	70700263	4395
1901			1928			1955	42355789	847	1982	70287711	4725
1902			1929			1956	44966245	1026	1983	67710786	5015
1903			1930			1957	45346843	1310	1984	72716099	4775
1904			1931			1958	42340660	1345	1985	71530126	4704
1905			1932			1959	44255923	1324	1986	70875075	4733
1906			1933			1960	45157969	1781	1987	71383246	4702
1907			1934			1961	45660632	1652	1988	72828956	4372
1908			1935			1962	47345229	1515	1989	73299825	4754
1909			1936			1963	49768279	1437	1990	74770215	4467
1910			1937			1964	51754508	1457	1991	74413311	4373
1911			1938			1965	53463287	1573	1992	73986513	4765
1912			1939			1966	56478580	1438	1993	72264608	4033
1913			1940			1967	59480253	1662	1994	74793769	4247
1914			1941			1968	61427650	1641	1995	75242995	4237
1915			1942			1969	63870798	1377	1996	76636826	4008
1916			1943			1970	66993845	1510	1997	76629473	4233
1917			1944			1971	66172590	1552	1998	77058186	5169
1918			1945			1972	67435669	1827	1999	75862260	5330
1919			1946			1973	67081951	1529	2000	75422096	5638
1920			1947			1974	65802709	1528	2001	75845604	5014
1921			1948			1975	64731176	1754	2002	74836815	6001
1922			1949	33466879	572	1976	64989908	2126	2003	74128040	6213
1923			1950	37495105	454	1977	65464382	2626	2004	74255012	6740
1924			1951	40881899	596	1978	66609727	2843	2005	73455026	6707
1925			1952	40002343	860	1979	69574945	2815	2006	74946029	7885
1926			1953	40280740	673	1980	70930247	3658	2007	75436328	10029
									2008	77765323	11429

¹³⁹ Activity represents United States production of total energy, defined at eia.doe.gov as "...fossil, nuclear and renewable power." Data is in billion Btu's as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ.

¹⁴⁰ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Fossil fuel, nuclear and renewables were used as the keywords found in the patent title or abstract by year of publication.

Table A4.24 Correlation Eq.(A1.1) terms calculated from Table A4.23 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
3.789E+09	201821	2.49E+17	1.01E+09	1.42E+13	9.363E+15	3.28E+08	1.41E+12	0.806353	65.02049

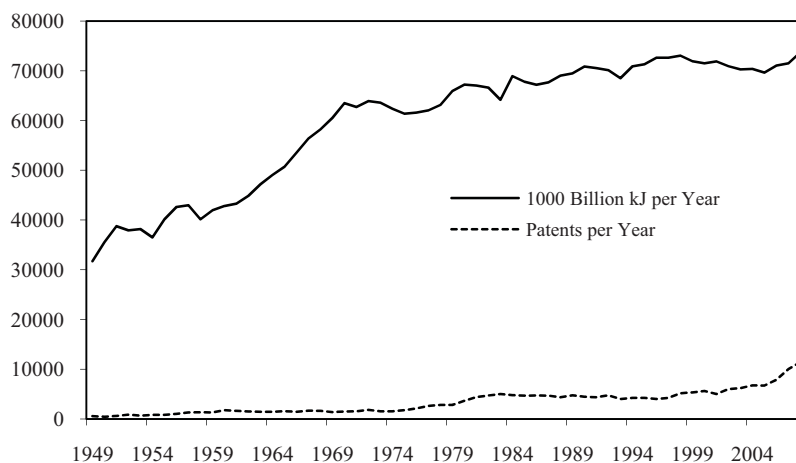


Figure A4.52. U.S. Total Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

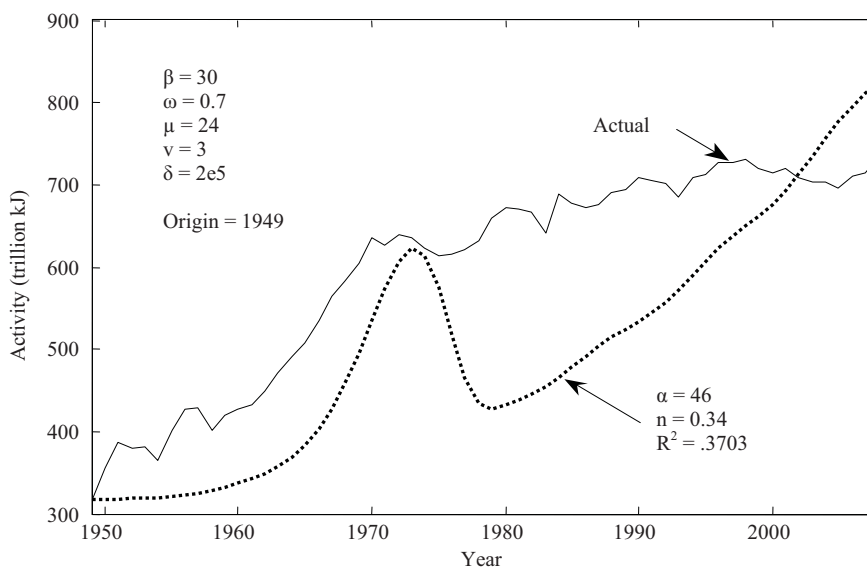


Figure A4.53. EIA U.S. Total Energy Production. U.S. total power production (activity) scaled in trillion kJ with actual and best-fit curves and common pattern equation parameters.

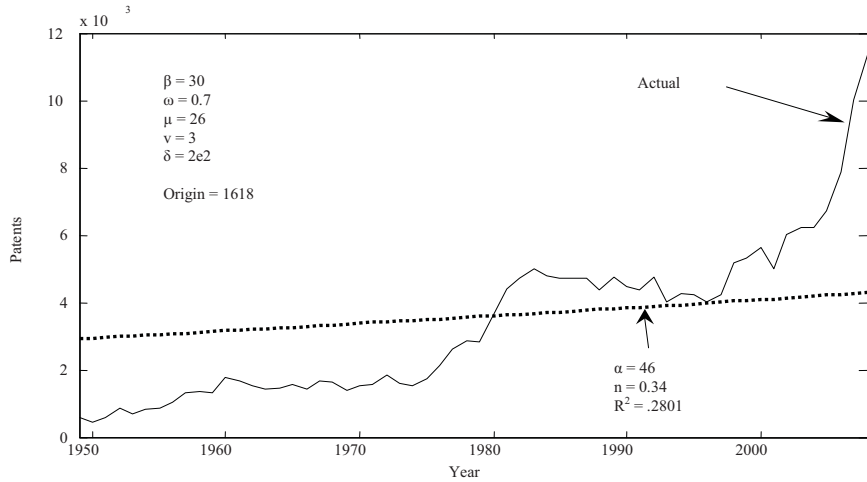


Figure A4.54. EPO Worldwide Patent Search: Fossil Fuel, Nuclear or Renewables in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

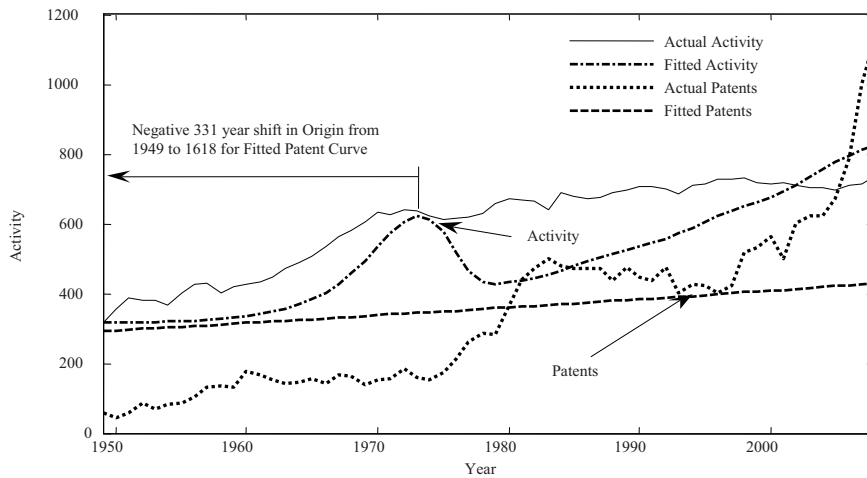


Figure A4.55. U.S. Total Energy Best-Fit Activity and Patents. Illustrates best-fit origin shift.

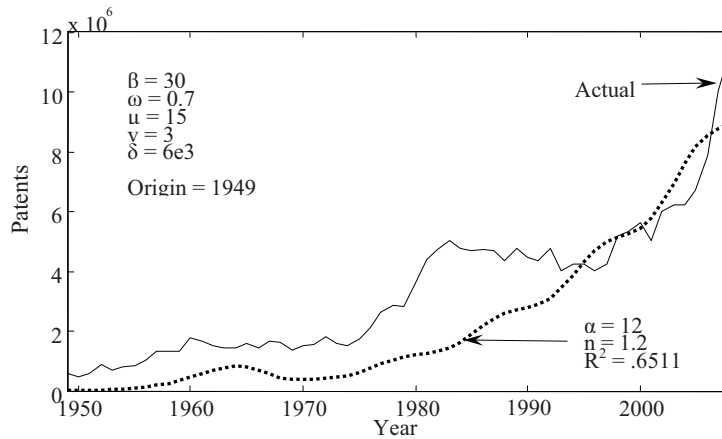


Figure A4.56. U.S. Total Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.27 U.S. Wind Power Activity¹⁴¹ and Patents¹⁴²

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954			1981		
1901			1928			1955			1982		
1902			1929			1956			1983	29.54	364
1903			1930			1957			1984	71.74	296
1904			1931			1958			1985	63.3	339
1905			1932			1959			1986	46.42	329
1906			1933			1960			1987	39.035	280
1907			1934			1961			1988	9.495	258
1908			1935			1962			1989	23244.82	310
1909			1936			1963			1990	30602.39	325
1910			1937			1964			1991	32489.78	301
1911			1938			1965			1992	31505.47	388
1912			1939			1966			1993	32691.29	376
1913			1940			1967			1994	37515.8	527
1914			1941			1968			1995	34424.65	543
1915			1942			1969			1996	35279.2	541
1916			1943			1970			1997	35427.96	601
1917			1944			1971			1998	32549.92	660
1918			1945			1972			1999	48418.17	913
1919			1946			1973			2000	60195.14	1002
1920			1947			1974			2001	73445.94	1161
1921			1948			1975			2002	111127.4	1420
1922			1949			1976			2003	120872.4	1773
1923			1950			1977			2004	149545.2	1912
1924			1951			1978			2005	187882.8	2053
1925			1952			1979			2006	278243.6	2357
1926			1953			1980			2007	359230.7	3182
									2008	542506.3	3805

Table A4.28. Correlation Eq.(A1.1) terms calculated from Table A4.27 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
2257458.4	26016	6.08E+11	49209118	5.25E+09	4.115E+11	23177108	2.99E+09	0.968074	93.71671

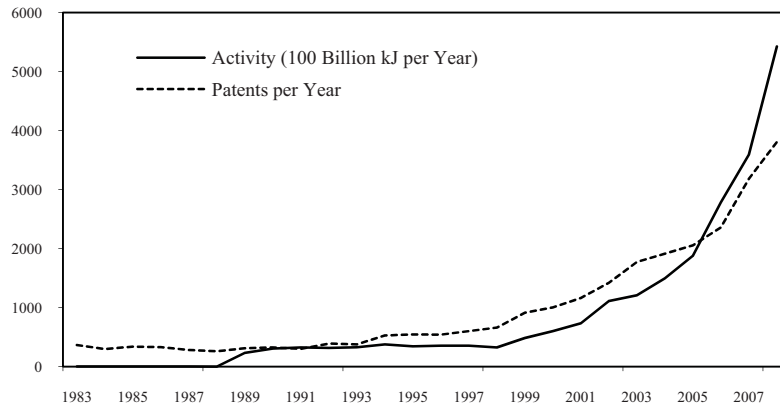


Figure A4.57. U.S. Wind Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

Activity represents U.S. production of wind energy, defined at usgs.gov as “wind electricity net generation (converted to Btu using the fossil-fueled plants heat rate).” Data is in billion Btu’s as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ.
¹⁴² Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Wind and (energy or power) were used as keywords found in the patent title or abstract by year of publication.

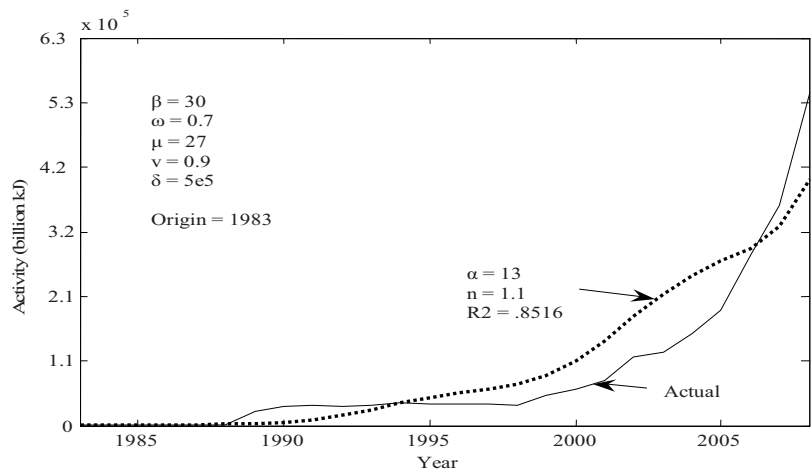


Figure A4.58. EIA U.S. Wind Energy Production. U.S. wind energy production (activity) scaled in billion kJ with actual and best-fit curves and common pattern equation parameters.

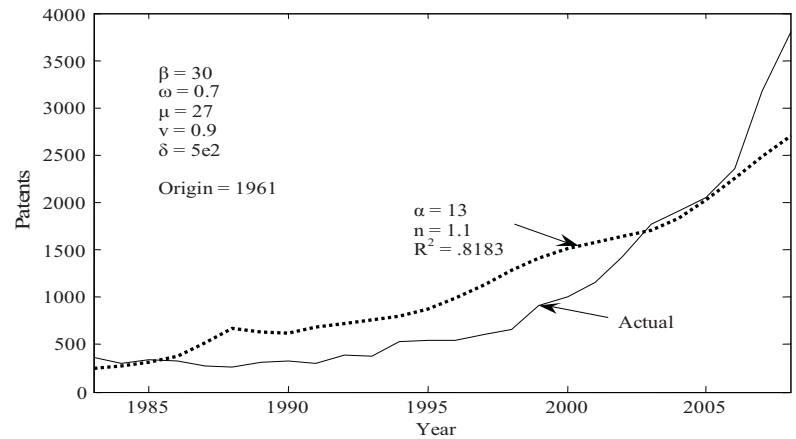


Figure A4.59. EPO Worldwide Patent Search: Wind and (Power or Energy) in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

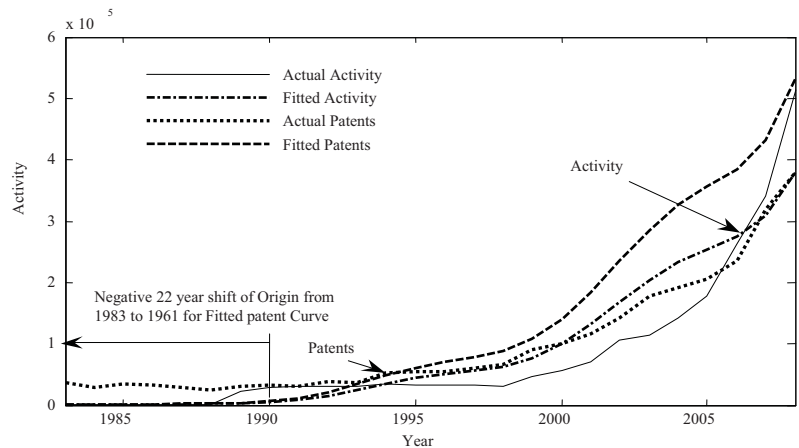


Figure A4.60. U.S. Wind Energy Best-Fit Activity and Patents. Illustrates best-fit origin shift.

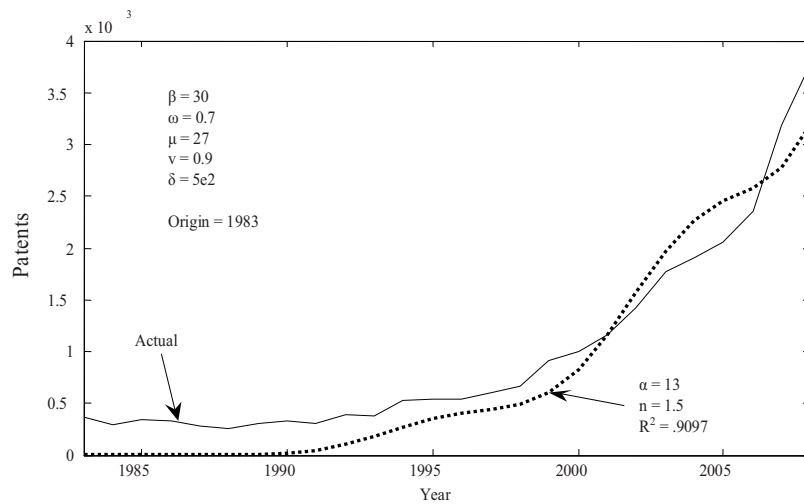


Figure A4.61. U.S. Wind Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A4.29 U.S. Wood Energy Activity¹⁴³ and Patents¹⁴⁴

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (pat)
1900			1927			1954	1471015	5	1981	2632819	70
1901			1928			1955	1502471	5	1982	2648101	80
1902			1929			1956	1493744	2	1983	2831906	79
1903			1930			1957	1406928	3	1984	2833537	88
1904			1931			1958	1395895	4	1985	2834537	77
1905			1932			1959	1427282	2	1986	2703051	70
1906			1933			1960	1392463	2	1987	2598633	55
1907			1934			1961	1365974	1	1988	2718379	44
1908			1935			1962	1371755	4	1989	2827002	45
1909			1936			1963	1396098	1	1990	2338054	52
1910			1937			1964	1410326	2	1991	2335858	51
1911			1938			1965	1408173	3	1992	2440712	48
1912			1939			1966	1444279	1	1993	2384062	49
1913			1940			1967	1413963	6	1994	2451630	60
1914			1941			1968	1497567	9	1995	2500212	57
1915			1942			1969	1519714	4	1996	2571063	60
1916			1943			1970	1507225	3	1997	2501396	52
1917			1944			1971	1508892	4	1998	2304289	66
1918			1945			1972	1583547	10	1999	2335946	81
1919			1946			1973	1610998	8	2000	2386109	86
1920			1947			1974	1622332	8	2001	2116154	93
1921			1948			1975	1579259	7	2002	2105024	68
1922			1949	1634471	4	1976	1805616	12	2003	2112152	75
1923			1950	1648234	2	1977	1937533	14	2004	2237920	86
1924			1951	1619076	6	1978	2148138	16	2005	2253850	82
1925			1952	1555459	6	1979	2268096	30	2006	2270076	85
1926			1953	1496624	4	1980	2609923	59	2007	2260250	88
									2008	2152850	120

¹⁴³ Activity represents United States production of wood power, defined at eia.doe.gov as energy from “Wood and wood-derived fuels.” Data is in billion Btu’s as reported by the Energy Information Administration (EIA) at eia.doe.gov converted to kJ.

¹⁴⁴ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Wood and combustion were used as keywords found in the patent title or abstract by year of publication.

Table A4.30 Correlation Eq.(A1.1) terms calculated from Table A4.29 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
119738640	2214	2.54E+14	152602	5.27E+09	1.491E+13	70905.4	8.5E+08	0.827053	68.40168

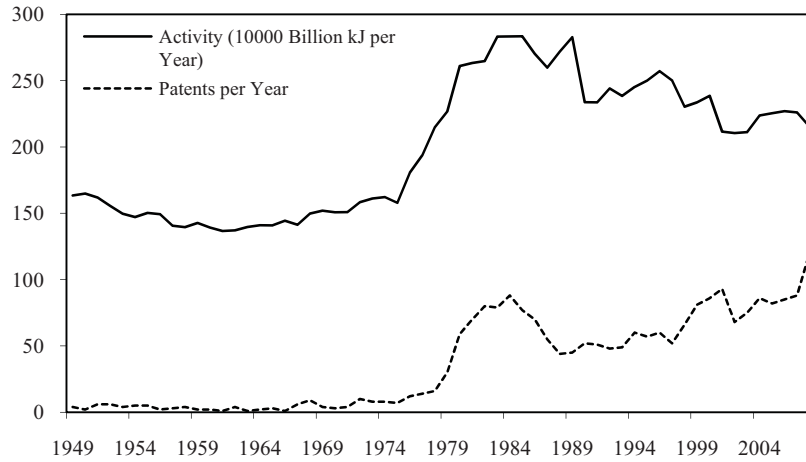


Figure A4.66. U.S. Wood Energy Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

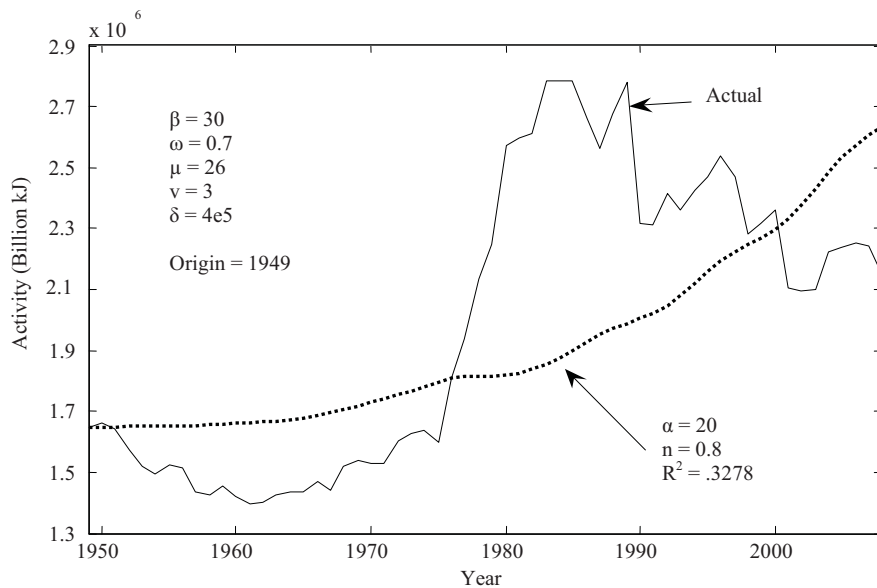


Figure A4.67. EIA U.S. Wood Energy Production. U.S. wood power (activity) production scaled in billion kJ with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

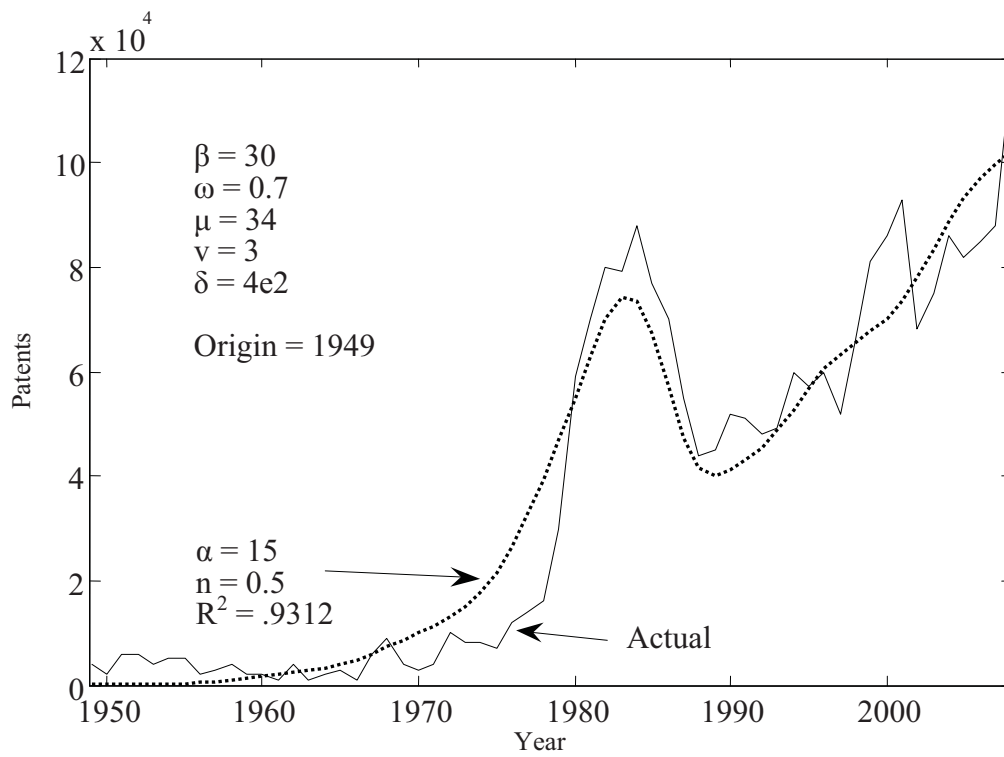


Figure A4.68. U.S. Wood Energy Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Appendix 5: Energy Materials Data

Table A5.1 U.S. Coal Activity¹⁴⁵ and Patents¹⁴⁶

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	268	125	1927	597	378	1954	421	267	1981	824	2289
1901	292	120	1928	575	440	1955	492	279	1982	838	2539
1902	300	138	1929	608	453	1956	532	312	1983	782	2675
1903	356	209	1930	536	513	1957	520	301	1984	896	2442
1904	350	210	1931	441	514	1958	433	232	1985	884	2266
1905	391	160	1932	359	382	1959	435	217	1986	890	2209
1906	412	163	1933	383	323	1960	436	285	1987	919	2200
1907	478	166	1934	417	373	1961	423	288	1988	946	1981
1908	414	165	1935	424	355	1962	441	262	1989	981	2022
1909	459	141	1936	493	343	1963	479	262	1990	1,029	1676
1910	500	202	1937	497	268	1964	506	270	1991	996	1842
1911	495	202	1938	394	312	1965	521	318	1992	998	1857
1912	533	176	1939	446	258	1966	549	281	1993	945	1556
1913	568	147	1940	512	189	1967	567	343	1994	1,034	1613
1914	511	142	1941	570	135	1968	559	327	1995	1,033	1561
1915	530	143	1942	643	117	1969	573	266	1996	1,064	1510
1916	589	98	1943	651	95	1970	615	297	1997	1,090	1596
1917	650	110	1944	683	113	1971	563	333	1998	1,118	1722
1918	677	140	1945	632	122	1972	603	373	1999	1,100	1781
1919	553	146	1946	594	123	1973	599	340	2000	1,074	1676
1920	657	230	1947	682	140	1974	610	356	2001	1,127.7	1660
1921	505	304	1948	657	182	1975	654	435	2002	1,094.3	1733
1922	476	364	1949	481	214	1976	685	648	2003	1,071.8	1843
1923	657	298	1950	560	162	1977	697	921	2004	1,112.1	2012
1924	571	319	1951	576	195	1978	670	1262	2005	1,131.5	2144
1925	581	361	1952	507	301	1979	778	1378	2006	1,162.7	3000
1926	657	306	1953	488	221	1980	830	1725	2007	1,146.6	4032
									2008	1,171.5	4778

Table A5.2. Correlation Eq.(A1.1) terms calculated from Table A5.1 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
71,481	85299	53066191	1.58E+08	75803600	6190218.4	91036847	19865741	0.836841	70.03036

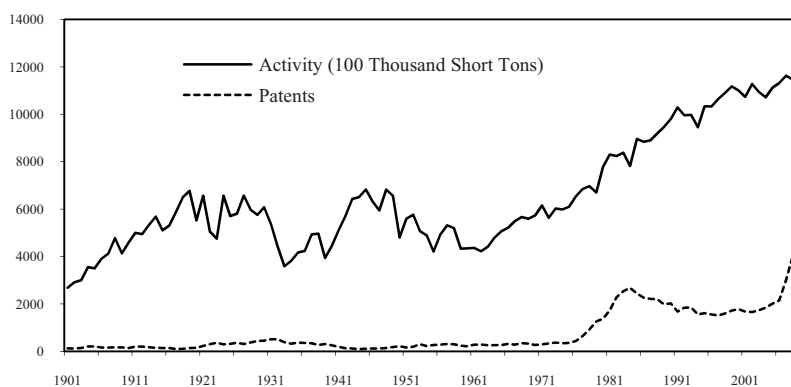


Figure A5.1. U.S. Coal Activity and Patents. data illustrates correlation. Activity scaled to fit plot.

¹⁴⁵ Activity represents United States production of coal, defined at eia.doe.gov as “Beginning in 2001, includes a small amount of refuse recovery” Data is in thousand of ktons as reported by the Energy Information Administration (EIA) at eia.doe.gov.

¹⁴⁶ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Coal was used as the keyword found in the patent title or abstract by year of publication.

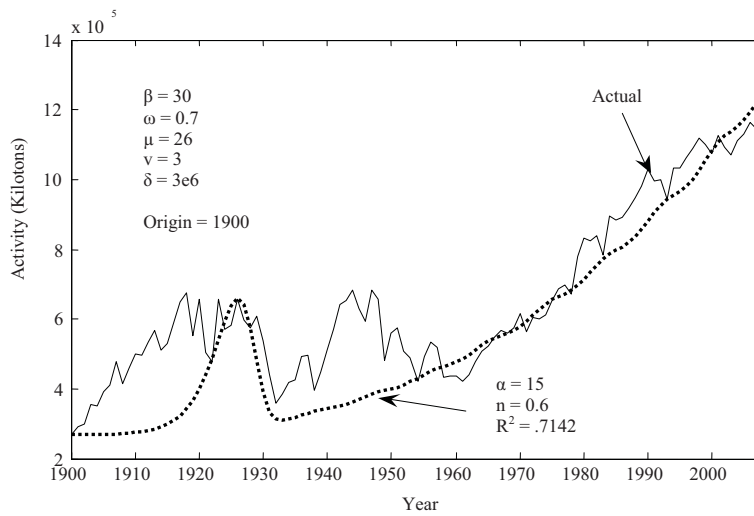


Figure A5.2. EIA U.S. Coal Production. U.S. coal production (activity) scaled in kilotons with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

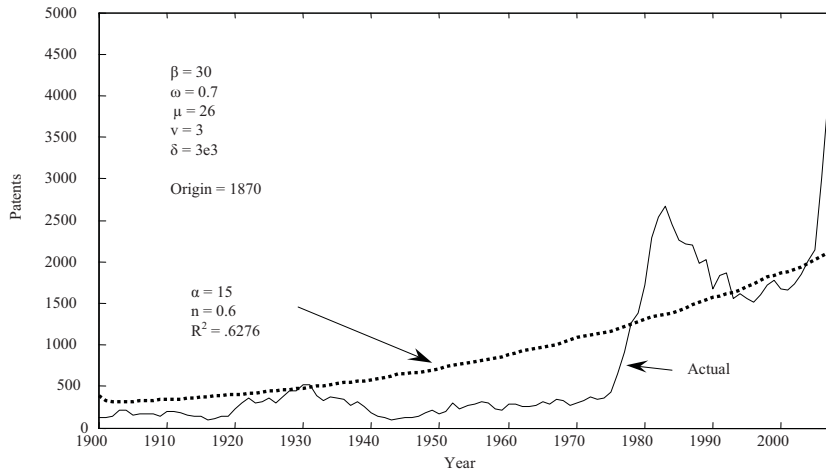


Figure A5.3. EPO Worldwide Patent Search: Coal in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

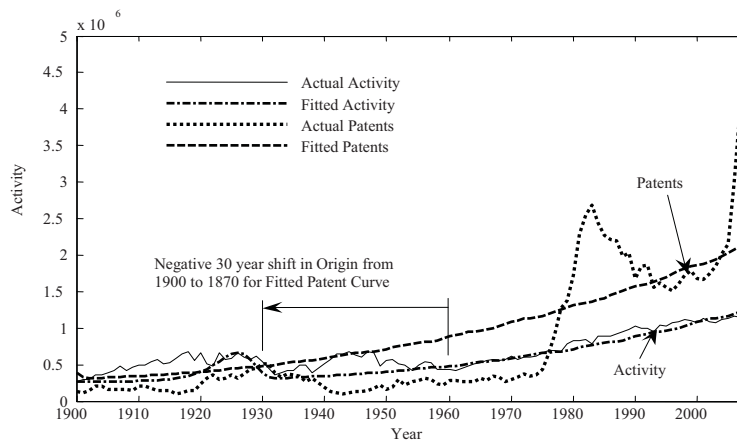


Figure A5.4. U.S. Coal Best-Fit Activity and Patents. Illustrates best-fit origin shift.

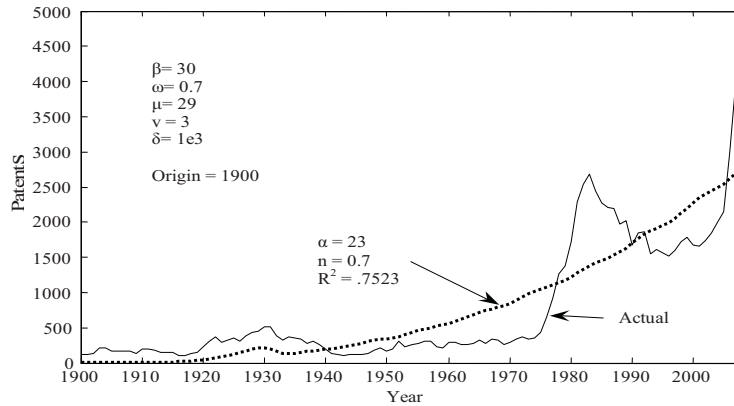


Figure A5.5. U.S. Coal Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A5.3. U.S. Natural Gas Activity¹⁴⁷ and Patents¹⁴⁸

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	10984850	210	1981	21587453	742
1901			1928			1955	11719794	223	1982	20272254	898
1902			1929			1956	12372905	254	1983	18659046	890
1903			1930			1957	12906669	301	1984	20266522	924
1904			1931			1958	13146635	306	1985	19606699	1015
1905			1932			1959	14229272	317	1986	19130711	1064
1906			1933			1960	15087911	452	1987	20140200	1017
1907			1934			1961	15460312	452	1988	20999255	1080
1908			1935			1962	16038973	408	1989	21074425	1204
1909			1936	2691512	120	1963	16973368	405	1990	21522622	1257
1910			1937	3084567	95	1964	17535553	463	1991	21750108	1209
1911			1938	3108858	119	1965	17963100	493	1992	22132249	1439
1912			1939	3387095	86	1966	19033839	424	1993	22725642	1263
1913			1940	3752702	65	1967	20251776	524	1994	23580706	1273
1914			1941	4168116	57	1968	21325000	516	1995	23743628	1310
1915			1942	4525095	51	1969	22679195	428	1996	24113536	1230
1916			1943	5024449	34	1970	23786453	446	1997	24212677	1254
1917			1944	5708288	45	1971	24088031	437	1998	24108128	1627
1918			1945	6000161	58	1972	24016109	594	1999	23822711	1736
1919			1946	6293037	88	1973	24067202	508	2000	24173875	2034
1920			1947	6733230	111	1974	22849793	456	2001	24500779	1986
1921			1948	7178777	127	1975	21103530	548	2002	23941279	2246
1922			1949	7546825	144	1976	20943778	581	2003	24118978	2322
1923			1950	8479650	112	1977	21097071	594	2004	23969678	2513
1924			1951	9689372	131	1978	21308815	535	2005	23456822	2515
1925			1952	10272566	213	1979	21883353	473	2006	23535018	2533
1926			1953	10645798	163	1980	21869692	744	2007	24590602	3133
									2008	26032337	3408

¹⁴⁷ Activity represents United States production of natural gas, defined at eia.doe.gov as “Natural gas gross withdrawals.” Data is in million cubic feet as reported by the Energy Information Administration (EIA) at eia.doe.gov.

¹⁴⁸ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Natural and gas were used as keywords found in the patent title or abstract by year of publication.

Table A5.4. Correlation Eq.(A1.1) terms calculated from Table A5. 3 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
1.245E+09	59033	2.5E+16	92636613	1.29E+12	3.794E+15	44898324	2.88E+11	0.698263	48.75718

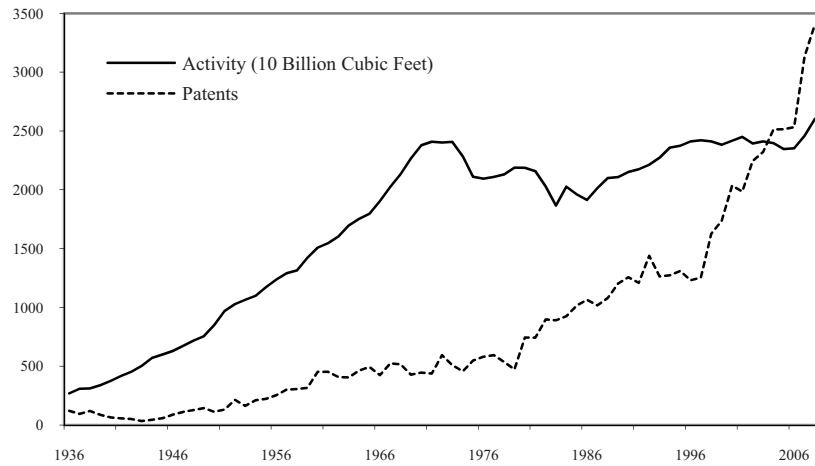


Figure A5.6. U.S. Natural Gas Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

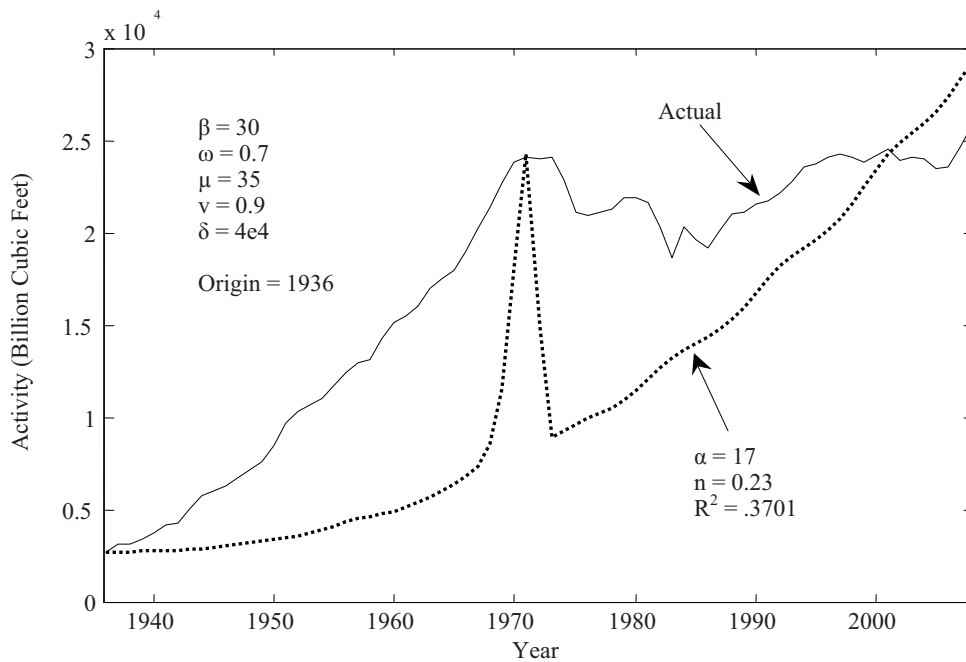


Figure A5.7. EIA U.S. Natural Gas Production. U.S. natural gas production (activity) scaled in billion cubic feet with actual and best-fit curves and common pattern equation parameters.

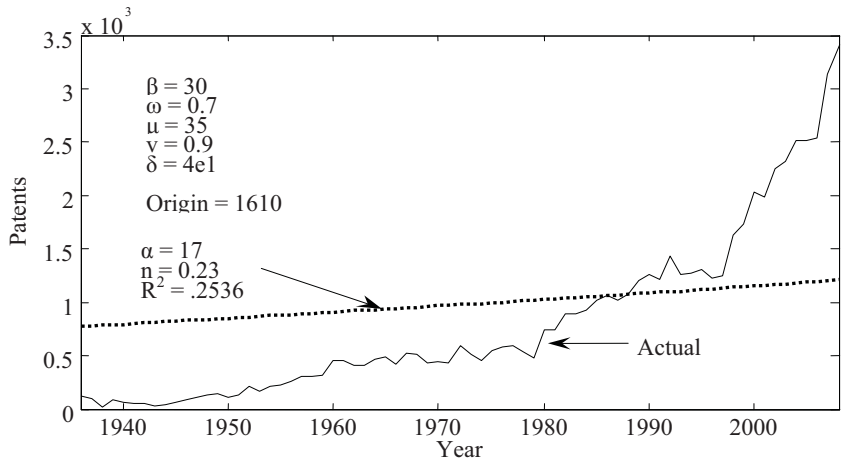


Figure A5.8. EPO Worldwide Patent Search: Natural Gas in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

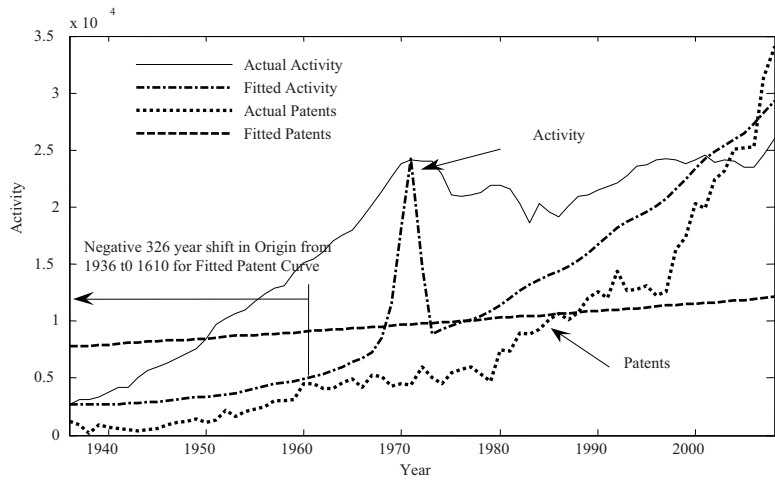


Figure A5.9. U.S. Natural Gas Best-Fit Activity and Patents. Illustrates best-fit origin shift.

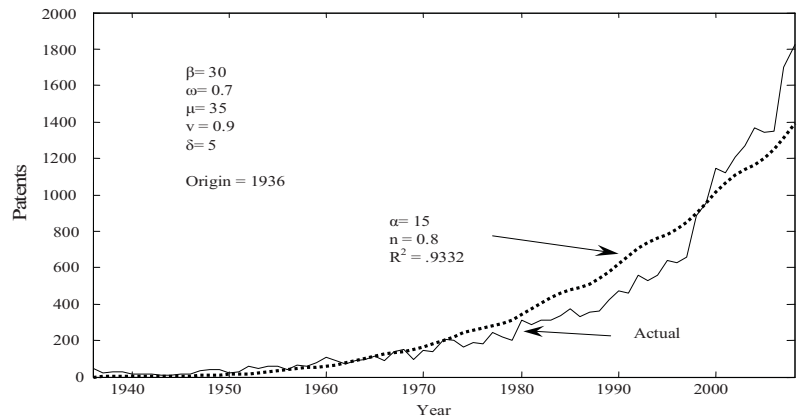


Figure A5.10. U.S. Natural Gas Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A5.5. U.S. Oil Activity¹⁴⁹ and Patents¹⁵⁰

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900	63621	55	1927	901129	109	1954	2314988	343	1981	3128624	807
1901	69389	60	1928	901474	146	1955	2484428	345	1982	3156715	873
1902	88767	56	1929	1007323	159	1956	2617283	460	1983	3170999	895
1903	100461	65	1930	898011	236	1957	2616901	558	1984	3249696	870
1904	117081	74	1931	851081	248	1958	2448987	537	1985	3274553	834
1905	134717	73	1932	785159	220	1959	2574590	471	1986	3168252	841
1906	126494	49	1933	905656	210	1960	2574933	575	1987	3047378	860
1907	166095	60	1934	908065	209	1961	2621758	482	1988	2979123	749
1908	178527	57	1935	993942	231	1962	2676189	400	1989	2778773	852
1909	183171	50	1936	1098513	248	1963	2752723	405	1990	2684687	936
1910	209557	47	1937	1277653	188	1964	2786822	355	1991	2707039	978
1911	220449	45	1938	1213254	222	1965	2848514	441	1992	2624632	920
1912	222935	51	1939	1264256	207	1966	3027763	424	1993	2499033	787
1913	248446	61	1940	1503176	179	1967	3215742	454	1994	2431476	883
1914	265763	55	1941	1404182	210	1968	3329042	426	1995	2394268	911
1915	281104	45	1942	1385479	149	1969	3371751	411	1996	2366017	859
1916	300767	26	1943	1505613	139	1970	3517450	450	1997	2354831	913
1917	335316	21	1944	1677904	153	1971	3453914	470	1998	2281919	1241
1918	355928	22	1945	1713655	147	1972	3455368	561	1999	2146732	1189
1919	378367	52	1946	1733424	154	1973	3360903	470	2000	2130707	1357
1920	442929	61	1947	1856987	195	1974	3202585	455	2001	2117511	1312
1921	472183	75	1948	2020185	232	1975	3056779	503	2002	2097124	1342
1922	557531	109	1949	1841940	214	1976	2976180	557	2003	2073453	1367
1923	732407	83	1950	1973574	180	1977	3009265	673	2004	1983302	1514
1924	713940	86	1951	2247711	270	1978	3178216	667	2005	1890106	1379
1925	620373	103	1952	2289836	346	1979	3121310	609	2006	1862259	1704
1926	770874	128	1953	2357082	289	1980	3146365	741	2007	1848450	1989
									2008	1811817	2190

Table A5.6. Correlation Eq.(A1.1) terms calculated from Table A5.5 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
196873681	51754	4.86E+14	46982414	1.23E+11	1.307E+14	22409235	2.92E+10	0.540152	29.17646

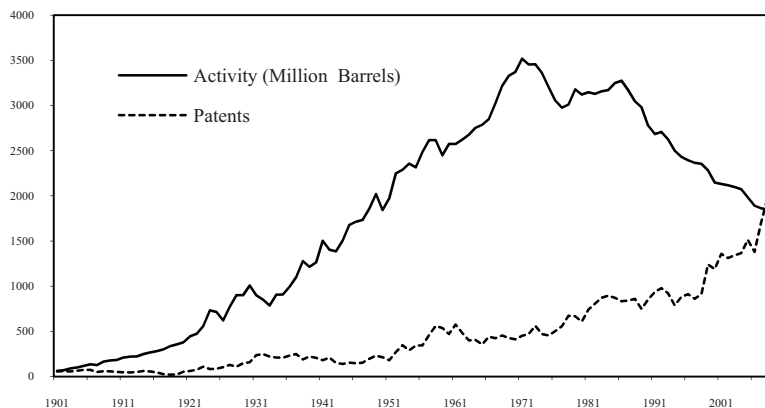


Figure A5.11. U.S. Oil Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

¹⁴⁹ Activity represents United States production of oil, defined at eia.doe.gov as “Field production of crude oil.” Data is in thousand barrels as reported by the Energy Information Administration (EIA) at eia.doe.gov.

¹⁵⁰ Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Petroleum was used as the keyword found in the patent title or abstract by year of publication.

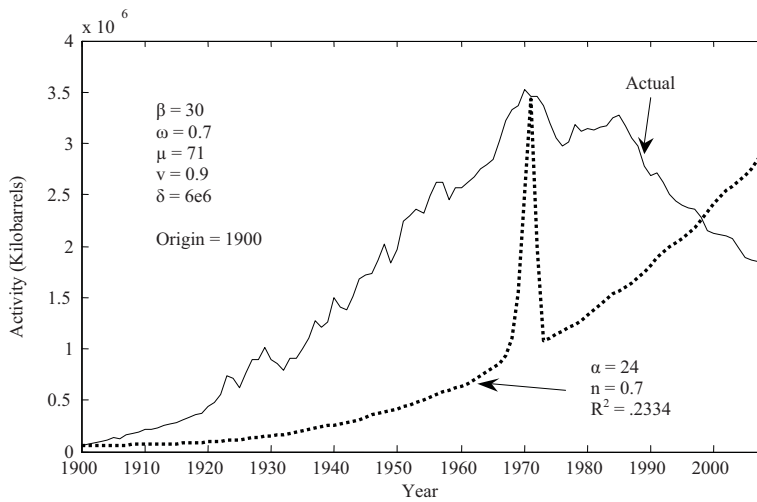


Figure 5.12. EIA U.S. Oil Production. U.S. oil production (activity) scaled in kilo barrels with actual and best-fit curves and common pattern equation parameters. No best-fit for the patent data was obtainable.

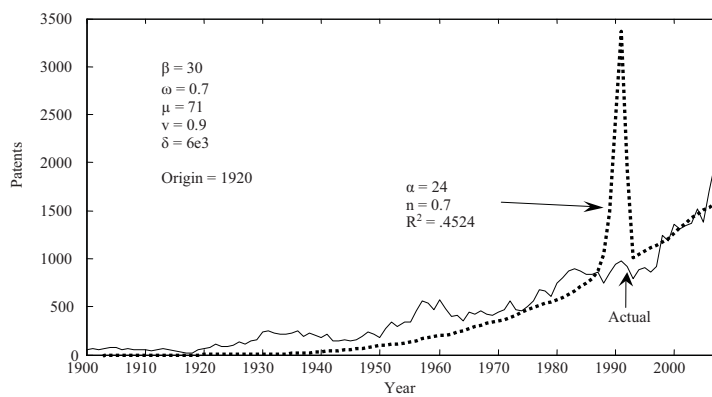


Figure A5.13. EPO Worldwide Patent Search: Petroleum in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

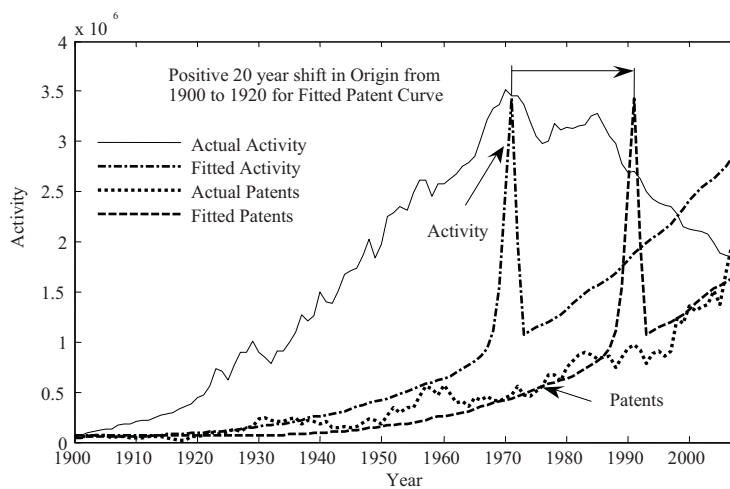


Figure A5.14. U.S. Oil best-Fit Activity and Patents. Illustrates best-fit origin shift.

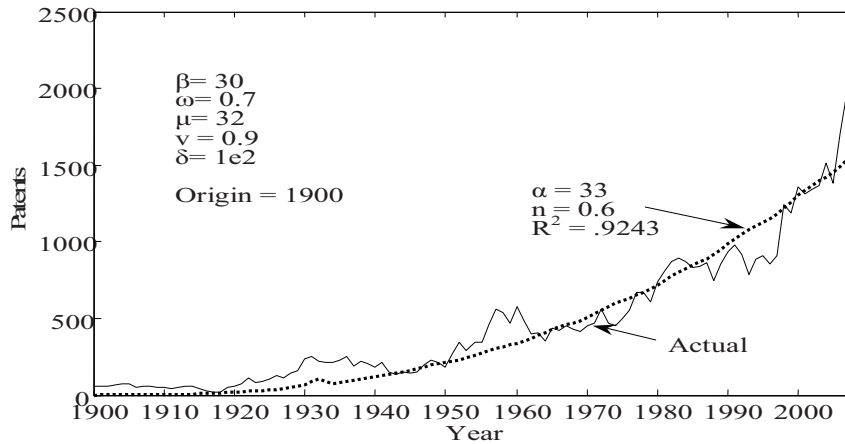


Figure A5.15. U.S. Oil Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Table A5.7. U.S. Uranium Usage Activity¹⁵¹ and Patents¹⁵²

Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)	Year	x (activity)	y (patent)
1900			1927			1954	4.95	43	1981	20.335	341
1901			1928			1955	6.58	43	1982	18.885	313
1902			1929			1956	12.21	79	1983	13.03	321
1903			1930			1957	17.03	131	1984	12.59	239
1904			1931			1958	28.59	236	1985	8.855	213
1905			1932			1959	34.39	271	1986	12.705	219
1906			1933			1960	35.64	306	1987	13.545	210
1907			1934			1961	31.85	234	1988	12.815	177
1908			1935			1962	29.11	196	1989	12.42	210
1909			1936			1963	25.42	199	1990	15.295	169
1910			1937			1964	17.9	200	1991	10.375	171
1911			1938			1965	14.44	182	1992	13.075	180
1912			1939			1966	12.49	184	1993	10.53	121
1913			1940			1967	10.55	184	1994	11.125	137
1914			1941			1968	11.57	134	1995	18.77	158
1915			1942			1969	11.11	107	1996	20.11	122
1916			1943			1970	10.805	125	1997	15.82	144
1917			1944			1971	17.075	135	1998	16.655	169
1918			1945			1972	12.8	130	1999	21.855	203
1919			1946			1973	12.635	140	2000	17.63	192
1920			1947			1974	10.03	137	2001	18.82	147
1921			1948			1975	11.8	135	2002	19.82	165
1922			1949	2.33	19	1976	13.945	172	2003	20.9	139
1923			1950	3.21	22	1977	15.74	229	2004	27.59	185
1924			1951	3.82	28	1978	17.685	213	2005	23.845	162
1925			1952	3.72	26	1979	17.135	233	2006	25.105	150
1926			1953	3.06	31	1980	20.75	271	2007	22.215	145
									2008	21.9	143

Activity represents U.S. uranium usage, defined at usgs.gov as domestic concentrate production and imports minus exports. Data is in thousands of tons as reported by the Energy Information Administration (EIA) at eia.doe.gov.

¹⁵² Patents are total patents by a worldwide data base patent search on the European Patent Office (EPO) search engine esp@cenet. Uranium was used as the keyword found in the patent title or abstract by year of publication.

Table A5.8. Correlation Eq.(A1.1) terms calculated from Table A5.7 data.

Sum x	Sum y	Sum x ²	Sum y ²	Sum xy	Sxx	Syy	Sxy	r	100r ²
956.985	10020	18620.23	1989320	180220.5	3356.5606	315980	20404.03	0.626526	39.25345

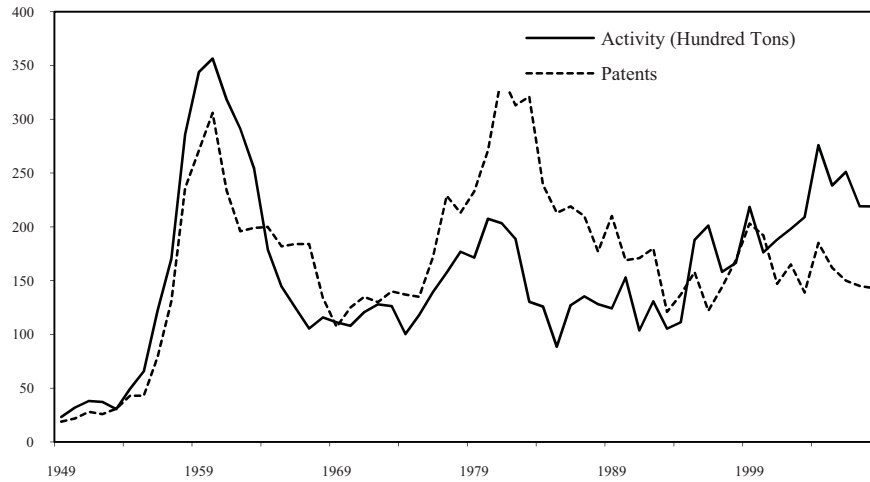


Figure A5.16. U.S. Uranium Usage Activity and Patents. Data illustrates correlation. Activity scaled to fit plot.

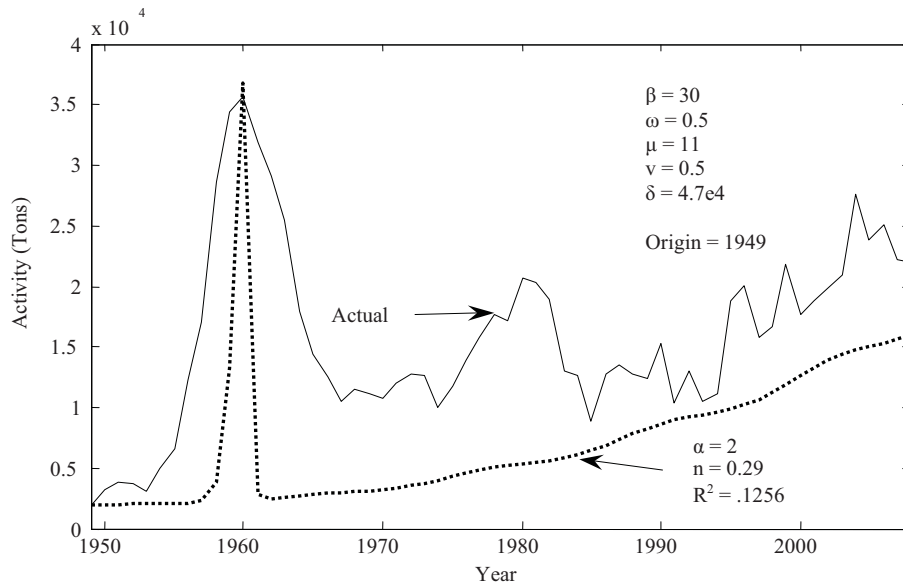


Figure A5.17. EIA Uranium Usage (Production and Imports minus Exports). U.S. uranium usage (activity) scaled in tons with actual and best-fit curves and common pattern equation parameters.

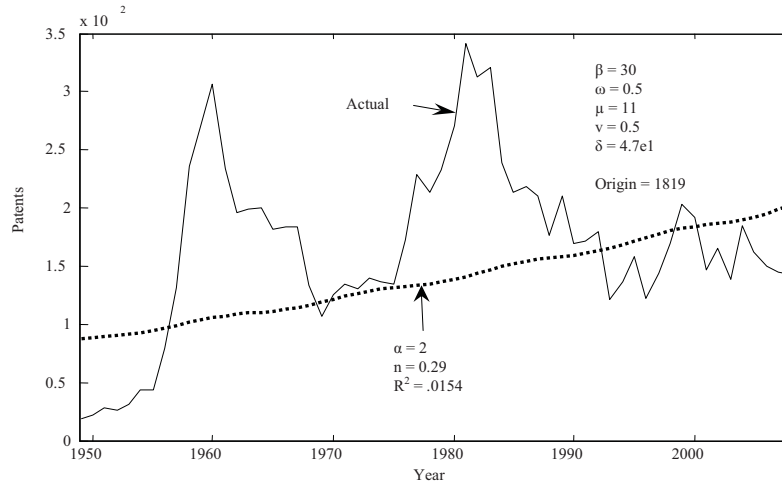


Figure A5.18. EPO Worldwide Patent Search: Uranium in Title or Abstract by Date of Publication. Best-fit generated using patent data in the production best-fit equation with production parameters. Only origin is changed.

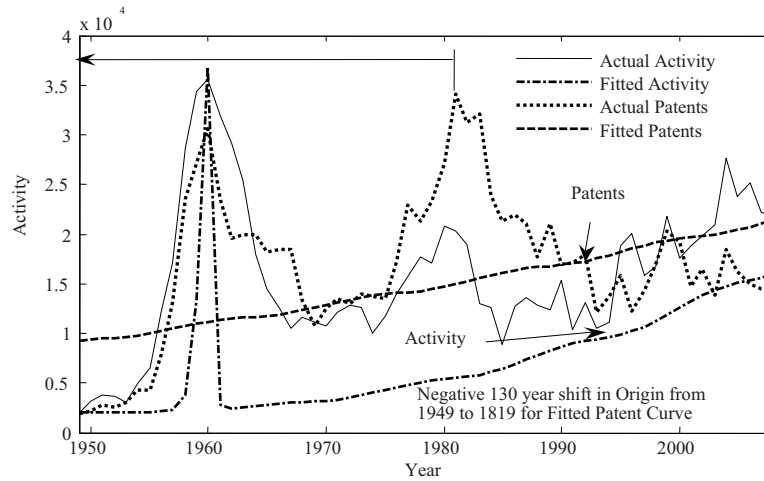


Figure A5.19. U.S. Uranium Best-Fit Activity and Patents. Illustrates best-fit origin shift.

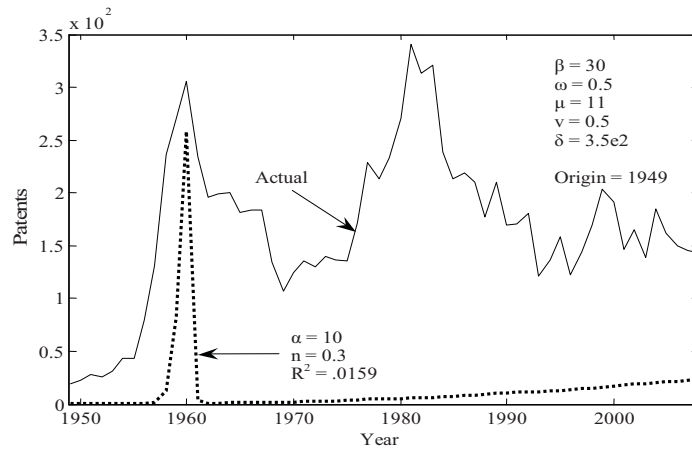


Figure A5.20. U.S. Uranium Independent Patent Best-Fit. Best-fit evaluation using patent data and pattern equation with unique patent equation parameters.

Appendix 6: Patent Search Keywords

Table A6.1. Engineering materials, energy sources and energy materials with the patent search keyword terms used for the patent search for each per year on the European Patent Office (EPO) Search Engine. Keywords were to be searched for in the title or abstract of the patents.

Engineering Material	Patent Search Keywords
Aluminum	Al, aluminum or aluminium
Antimony	antimony or stibium
Arsenic	arsenic
Asbestos	asbestos
Barite	barite, baryte or (barium and sulfate)
Bauxite/Alumina	bauxite or alumina
Beryllium	beryllium
Bismuth	bismuth
Boron	boron
Cadmium	cadmium or Cd
Chromium	cobalt
Cobalt	chromium, chrome or Cr
Copper	copper or Cu
Feldspar	feldspar
Fluorspar	fluorspar or fluorite
Gold	gold
Graphite	graphite
Gypsum	gypsum
Helium	helium
Hydraulic Cement	hydraulic and cement
Iodine	iodine
Iron	iron or Fe
Kyanite	kyanite or (aluminum and silicate)
Lead	lead or Pb
Lithium	lithium or Li
Magnesite	magnesium and carbonate
Magnesium	magnesium or Mg
Manganese	manganese or Mn
Mercury	mercury or Hg
Molybdenum	molybdenum or Mo
Nickel	nickel or Ni
Niobium	niobium, Nb or columbium
Nitrogen	nitrogen
Phosphate Rock	phosphate
Platinum	platinum or Pt
Potash	potash
Rare Earths	lanthanide, lanthanoid, yttrium or scandium
Salt	salt
Selenium	selenium
Silicon	silicon or Si
Silver	silver or Ag
Sulfur	sulfur or sulphur

Talc	talc or pyrophyllite
Tantalum	tantalum or Ta
Tin	tin or Sn
Titanium	titanium or Ti
Tungsten	tungsten or wolfram
Vanadium	vanadium or V
Zinc	zinc or Zn
Zirconium	zirconium or Zr
Energy Sources	Patent Search Keywords
U.S. Biofuel Energy	biofuel or biofuels or biodiesel
U.S. Biomass Energy	biomass
U.S. Coal Energy	coal
U.S. Fossil Fuel Energy	coal + (natural and gas) or methane or ethane + petroleum
U.S. Geothermal Energy	geothermal and (power or energy)
U.S. Hydroelectric Energy	hydroelectric
U.S. Natural Gas Energy	(natural and gas) or methane or ethane
U.S. Nuclear Energy	(nuclear and(power or energy)) or uranium
U.S. Oil Energy	petroleum
U.S. Renewable Energy	renewable and energy
U.S. Solar Energy	solar
U.S. Total Energy	coal + (natural and gas) or methane or ethane + petroleum + renewable and energy + (nuclear and(power or energy)) or uranium
U.S. Wind Energy	wind and (energy or power)
U.S. Wood Energy	wood and combustion
Energy Materials	Patent Search Keywords
Coal	coal
Natural Gas	natural and gas
Oil	petroleum
Uranium	uranium

Appendix 7: Scaling

In general, the production data is entered in the best-fit production equation as thousands of tons. For many materials this leads to a high R^2 value and also to an eventual shift in origin. In other cases the production must be scaled differently to achieve a high R^2 value and an origin shift. Scaling is sought to achieve the least differential between the scale of the production and patent data. A material with production numbers much larger than its patent numbers may need to be scaled up and entered as thousands of kilotons or megatons. When patent data counts are much greater than production, the production is entered as thousands of kilograms. This procedure allows for a more accurate evaluation of the best-fit of both production and patent data by resulting in R^2 values generally closer to one and resulting origin in shifts, but also is representative of the same amount of production only in a different scale.

Table A7.1 lists the materials which were tested with their production in tons and scaled in kilotons, megatons or kilograms. Listed also are three energy sources that were evaluated in billion kJ and scaled in trillion kJ. Some patterns may be revealed. In many cases R^2 values are closer to one for the scaled production best-fits meaning the fit is better in these cases. When the R^2 value is less for the scaled data it is still near the original R^2 . When a material or source is scaled up the numbers entered into the Matlab program are smaller by units of a thousand (kiloton, megaton, trillion kJ) depending on the new scale. For example, one million tons would be entered in the program as 1000. But when scaled in kilotons it is entered as one. The resulting y-axis data would be less by a degree of 1000 as would the δ parameter (e.g. $4e6$ goes to $4e3$) which is the peak amplitude of the Stage II hump given in tons. In all cases δ decreased by the same multiple that the entered scaled data was subjected to.

Similarly when a material or source is scaled down the numbers entered into the Matlab program are larger by units of a thousand (kilogram) depending on the new scale. For example, one ton would be entered in the program as .001. But when scaled in kilograms it is entered as one. The resulting y-axis data would be more by a degree of 1000 as would the δ parameter (e.g. 4e3 to 4e6) which is the peak amplitude of the Stage II hump given in tons. In the only case of scaling down presented here, δ decreased by the same multiple that the entered scaled data was subjected to. Platinum had no original parameters to compare because definite ones are not produced with a negative R^2 .

Table A7.1. Materials and energy sources that had scaled production data to achieve a modified patent data origin shift. Production is scaled up or down to better match the scale of the patent data. Platinum had no α , n , μ or δ for material in tons since with the case of a negative R^2 these best value of these parameters cannot be precisely determined. Best-fit parameters not listed (β , ω and ν) did not change when materials or sources were scaled. δ may not have change exactly the by the same multiple as the scaling, but this is due to the original in tons not being at the optimal amount.

Engineering Material in Tons	α	n	δ	μ	R^2	Units Scaled to	α	n	δ	μ	R^2
Barite	14	1	1e6	11	.5876	Ktons	14	.017	1e3	11	.7803
Bauxite/Alumina	22	1.25	8e6	43	.9776	Ktons	59	0.3	4e4	43	.9521
Feldspar	13	1	1e6	19	.8925	Ktons	11	.01	1e3	19	.7833
Fluorspar	18	0.9	0.6e6	30	.6347	Ktons	1	.003	.6e3	30	.5703
Gypsum	29	1.29	4e7	39	.7524	Ktons	28	0.4	4e4	6	.8740
Hydraulic Cem.	16	1.8	1e8	16	.9224	Mtons	15	.05	1e4	16	.9743
Iron	18	1.6	1.3e8	25	.7549	Ktons	13	0.7	1.3e5	25	.8599
Magnesite	22	1	1e6	30	.6706	Ktons	35	.08	1e3	30	.8231
Nitrogen	14	1.58	1e7	15	.5337	Ktons	14	.56	1e4	15	.7316
Phosphate	33	1.2	5e6	30	.7896	Ktons	30	.35	5e3	30	.8815
Platinum	-	-	-	-	Neg.	Kilograms	14	0.5	1e4	10	.9539
Salt	19	1.4	5e6	7	.6760	Ktons	19	.47	5e3	7	.8438
Silver	4	0.1	1.5e5	11	.6964	Kilograms	5	1.2	1.5e8	11	.5027
Sulfur	16	1.2	5e6	3	.8399	Ktons	20	.25	5e3	3	.9322
Talc	11	1	1e6	39	.8551	Ktons	1	.01	1e3	39	.9226
Energy Sources in Billion kJ	α	n	δ	μ	R^2	Units Scaled to	α	n	δ	μ	R^2
Coal	18	0.6	4.2e6	26	.2633	Trillion kJ	28	0.2	4e3	26	.8547
Fossil Fuel	24	1.2	8.4e7	26	.0153	Trillion kJ	36	0.29	2e5	23	.1995
Total Energy	20	1.3	4.2e5	26	.0258	Trillion kJ	46	0.34	2e5	24	.3703

It is possible that the relative changes from year to year in the production data create the features in the plots that determine the stage of the material rather than the scale of the data. A

material would be in the same stage whether its data is in tens, hundreds or thousands because the relative changes between data points would be the same from year to year resulting in identical plots of the actual production data with different y-axis scales. As shown in Figs. A7.1 and A7.2 there is a change in the fitted curve when the production is scaled to kilotons. This change would be caused by the differing best-fit parameters used in either case.

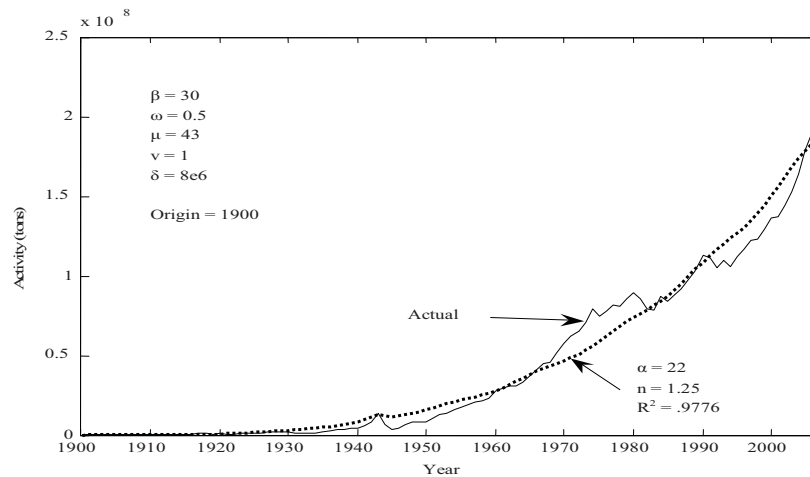


Figure A7.1. Bauxite production best-fit with production in tons.

For example, best-fit parameters were determined for bauxite as shown by Fig. A7.1 using tons as the unit for production activity producing a strong R^2 of 0.9776. However the numbers entered for the production data were in the tens of thousands to the hundreds of Millions while the entered patent data ranged from the tens to the thousands. The δ peak, which is measured in tons, produced by the production was too large to allow the patent data to run with the production equation. The δ peak was larger than the patent data itself. By scaling up to kilotons the entered production data was in the range of tens to hundreds of thousands with the patent data remaining the same. This scaling factor was mirrored by the δ value and peak in Fig. A7.2 that resulted in a scale that could be handled by the patent data. The scale also matches more accurately the scale of the production data in the y-axis. The shift in the production data of

a certain multiple of 1000 is matched by the δ being scaled by the same amount allowing the data set and equation to use the same y_0 as the un-scaled one, leaving α dimensionless and rendering such scaling permissible. The same procedures were performed for all of the scaled engineering materials of energy sources.

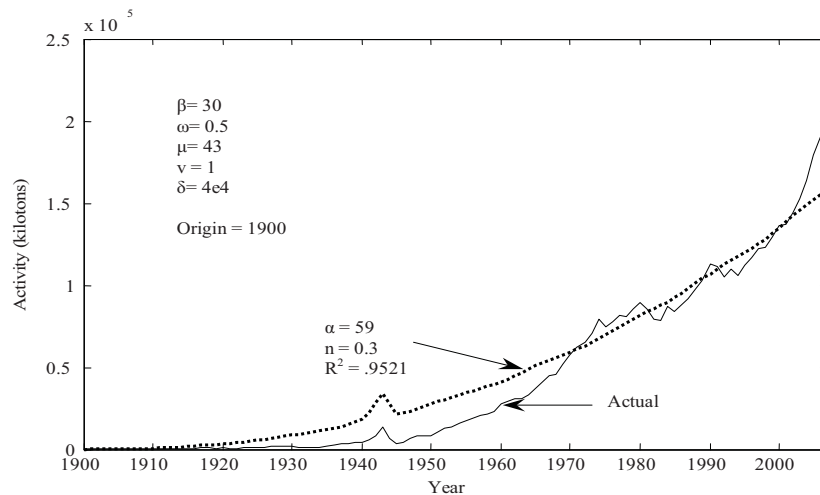


Figure A7.2. Bauxite production best-fit plot with production scaled in metric kilotons. The actual patent data curve is identical in shape to the curve in Fig. A6.1. Note the fitted curve is different due to different parameters.

While the δ parameter used for the calculation of the modified best-fit was identical to the δ used to obtain the production best-fit, the δ presented on all of the modified and independent patent best-fit figures was scaled down by a factor of one thousand. If the δ was $1e6$ for production it was recorded on the figure as $1e3$ since this better reflects the scale of the entered patent data. The patent data was not entered as one thousands as was the production data, but was still processed by an equation that treated the data as one thousands. The resulting δ values would describe a Stage II hump that was greater than the scale of the data itself. Scaling back δ solved this discrepancy with no change in results and only in the scale of the Stage II hump to reflect it more accurately.

Appendix 8: Executive Summary

- 1) Long-term life cycles can be used to describe the patterns that exist in the overall lifetime of a product, system or even living organism [2,84-89]. Such cycles exist in nature and in the business, manufacturing and engineering worlds. Life cycles may display the changes in the production of an item or the physical growth of a system or organism over a period of years. Four common stages have been identified in the past within the life cycle allowing for an evaluation of the trends that are present in the item and giving an indication of where the item is positioned in reference to its life cycle.
 - a) Stage I: This stage is called the Initial Stage and is a developmental stage that begins with the discovery and invention of a process or product and ends when the development of the technology is great enough to start low-scale industrial production [2].
 - b) Stage II: Stage II is known as the Lift Off and Decay Stage and begins with the rise of in the activity, or production, of a material and ends at the low point, or “valley of death” of the activity. Invention driven activity occurs in Stages I and II [2]
 - c) Stage III: Stage III, or Revival and Rapid Growth starts at the “valley of death” and continues through the material’s full growth potential with the take-off in activity typically being at a high rate [2].
 - d) Stage IV: This stage is called the Survival or Low Growth Stage and is where the material has reached maturity and the activity has leveled off or has begun to die. Innovation is dominant in Stages III and IV [2]

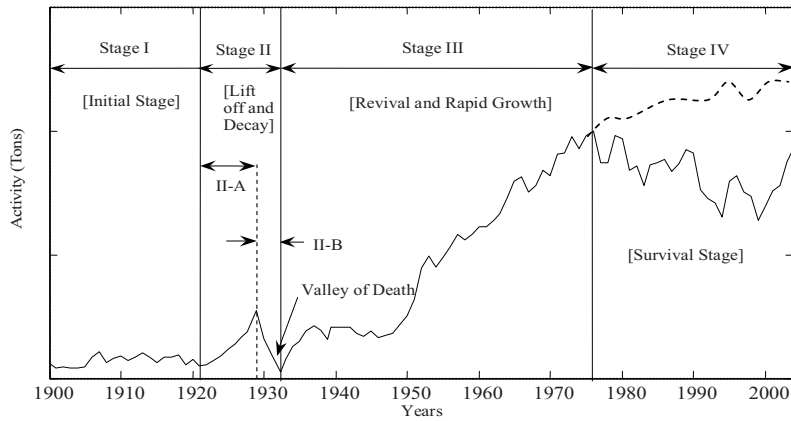


Figure A8.1. Illustration of a Typical Long-term Life Cycle for a Material. The plot indicates the division of the life cycle into four stages that is common to metals and non-metals. Common stage features are displayed.

2) The features common to the various life cycle stages are presented in Fig. A8.1. These features may be obvious from the raw data, but at times the stage of a material or system is not obvious. A model was needed that would mathematically capture the features of the stages of the life cycle and present a fitted curve that would represent the life cycle of a material in a form more suitable for evaluation.

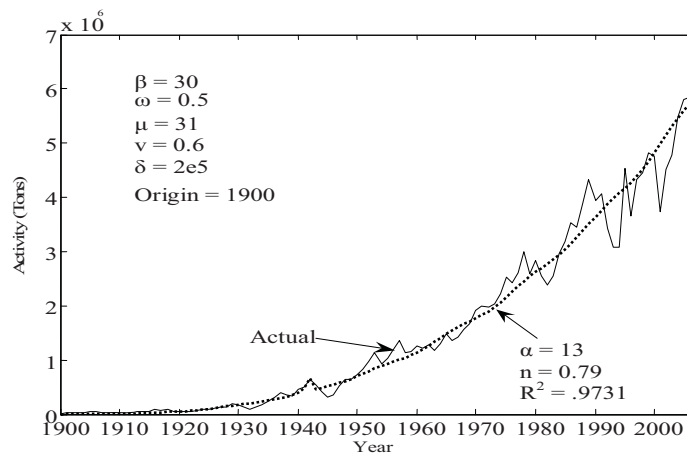


Figure A8.2. USGS World Chromium Production. Typical fitted production activity curve with best-fit parameters and R^2 value and origin. Shown in the figure are both the actual data curve and a best-fit curve.

- a) A life cycle model for metals was proposed by Yeramilli in 2006. An equation, $y = x^n [\alpha x^2 + \beta x \sin(\omega x)] + (\exp[(x - \mu) / \nu] \exp[-\exp[(x - \mu) / \nu]] \delta / \nu)$, was developed that revealed that metals could be modeled successfully and that they displayed four-stage life cycles [2]. Such a model is shown in Fig. A8.2.
- b) This equation was modified by Connelly in 2008 by raising the α and β parameters to the power of n leading to more consistent and reliable results [1].
- c) Here, the equation is further modified by the normalizing of n by dividing it by n_0 which is equal to one, thereby making n dimensionless.
- d) Also modified here to account for early years of production data which are far above zero and, as such, are much greater than the corresponding fitted curves resulting in low R^2 values. A constant, called CI , equal to the first year of the production data is added to the equation leading to $y = CI + x^n [\alpha^n x^2 + \beta^n x \sin(\omega x)] + (\exp[(x - \mu) / \nu] \exp[-\exp[(x - \mu) / \nu]] \delta / \nu)$ (2). Employed only with iodine, lead and silicon where more reasonable R^2 resulted.
- e) The first part of the equation, $x^n [\alpha^n x^2 + \beta^n x \sin(\omega x)]$, is responsible for the shape of Stage III (and Stage IV if a parabolic Alpha is used) in the fitted curve while the second part, $+ (\exp[(x - \mu) / \nu] \exp[-\exp[(x - \mu) / \nu]] \delta / \nu)$, controls the shape of Stages I-II in the fitted curve.
- i) Therefore, the second part of the equation is concerned with invention and the first part of the equation is concerned with innovation.

ii) Alpha and n are found only in the first part of the equation and only affect the innovative portion of the life cycle (Stage III). If alpha is small, the inventive stages are stretched out, are longer, and dominate the fitted curve. Conversely, if alpha is large, then the inventive stages are shorter and the innovative stage is longer and dominates.

- (1) Large alphas could indicate that a material is more mature and is farther into Stage III, or the innovative portion of its life cycle. Small alphas may indicate that a material is just entering Stage III and less advanced in its overall life cycle.
 - (2) Alpha may be a measure of initial invention, measuring the period of time that was needed to develop the material inventively.
 - (3) Since alpha, beta and x are all raised to the power n , n may be a measure of the importance of an invention. Materials such as talc with small n values (0.01), might then be of low technical importance and approaching Stage IV along Path A in Fig. A8.5.
- 3) This modified equation was applied to obtain a best-fit for the production, in tons per year, for over fifty engineering materials (metals and non-metals) and also for their associated patents.
- a) The production data and patent data were evaluated separately to reveal fitted curves modeling life cycle stages I-III. Linear alphas were used in the common pattern equation to reveal Stage III behavior. (Parabolic alphas model and reveal Stage IV, but were not employed here since only Stage III materials were sought.)

- b) The result of each production and patent data best-fit evaluation was a fitted curve and an R^2 value indicating the extent of the fit of the generated curves. An R^2 value near one for the fitted curve for the production of a material was considered as an indication that the material was in Stage III of its production life cycle. Likewise, an R^2 value near one for the fitted curve for the patent data of a material was considered as evidence that the material was in Stage III in its patent data life cycle.
- c) Correlation theory was applied to the production and the patent data to explore if any relationships exist between the two data sets for each material. It was found that the production data and patent data for materials were correlated to some degree in most cases across their entire life cycles. This implies that changes in the production data set can be attributed to changes in the patent data set or vice versa. If strong correlation exists, Stage III is assumed to be present. Stage IV or V (Stage V being a newly introduced stage beyond Stage IV where production is in a very steep decline with little evidence of future recovery), is assumed if correlation is very weak or is not indicated.
- 4) The point at which a material leaves Stage III and enters Stage IV is an important concept that, as of yet, has no clear answer. This transition from Stages III to IV could take place over a span of several years. The transition could occur due to resource depletion, environmental concerns or low impact of new patents. A parabolic alpha, when employed in the equation presented here, may be useful in describing a Stage IV material. Conversely, Stage III may be indicated when a constant, or linear alpha, produces a fitted curve with a good R^2 value. Further best-fit analysis presented here may offer a possible dividing line between Stage III and Stage IV.

- a) The patent data was manipulated to reveal a new best-fit curve with a new origin that can be either earlier or later than the original.
- i) The patent data was entered into the equation used to determine the best-fit for the production data. The data set was merged by using dimensions of the corresponding data set. All parameters were kept the same except for the data set and the data origin.
 - ii) The origin was move backward or forward to obtain the R^2 value nearest to one. This change in origin became a positive (forward) or negative (backward) origin shift.
 - iii) The value of R^2 is sensitive to the movement of the origin leading to an optimal R^2 value. A change in the year away from the origin shift that produces the R^2 nearest one results in a typical steep drop-off away from the optimal R^2 . Figure A8.3 shows that for biofuel energy the R^2 nearest one was obtained with an origin shift of seven years from 1981 to 1988. The R^2 values drop steeply away from this value as the origin shift moves away from the origin shift of seven years in both directions. A best value of R^2 is thereby assured in all cases.

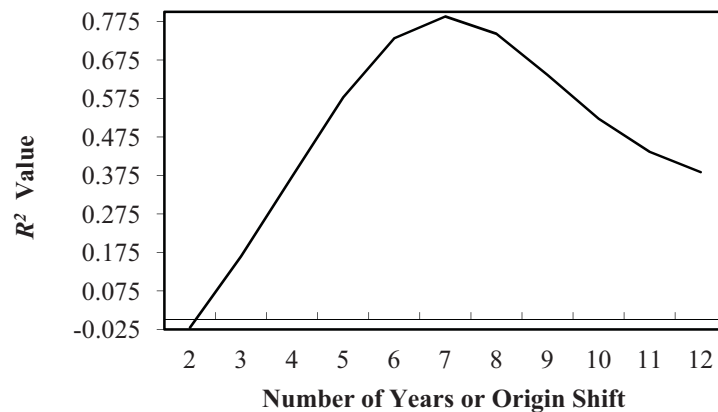


Figure A8.3. Biofuel Energy Sensitivity Curve: R^2 Values by Number of Years Shifted From Origin. Origin at 1981, R^2 closest to 1 at 1988. Illustrates steep drop-off of R^2 after best R^2 was attained at a positive seven year shift in origin. Such drops are typical for all origin shift evaluations.

b) This new curve can be superimposed upon the original to graphically reveal a shift in origin. A typical origin shift is shown for Zinc in Fig. A8.4.

i) Shifts may be seen by examining the Stage II humps if they are prominent. They often are not obvious. Shifts may be seen in both the actual and fitted curves and are the same for both.

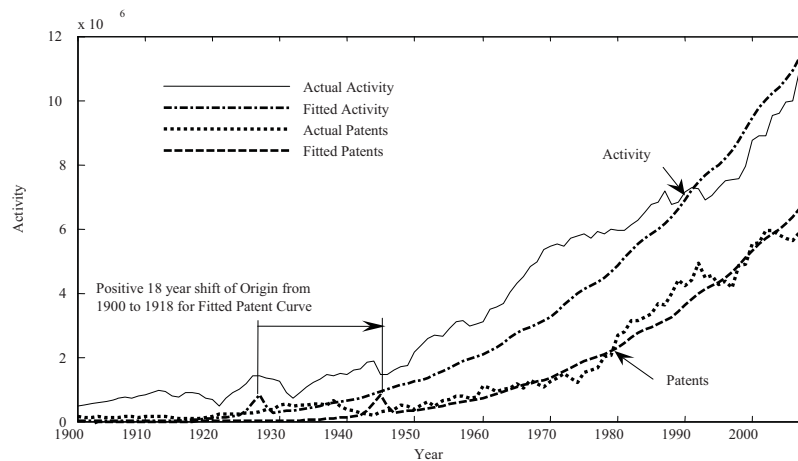


Figure A8.4. Zinc Best-Fit Activity and Patents. Plot depicting the origin shift of patent and activity best-fit curves for zinc. The shift is positive, indicating patent activity occurring after production activity and thus possibly being driven by the production. All parameters for the pattern equation are identical for the patent and production activity curves except for the origins (the matching results in the positive origin shift).

ii) A positive shift occurs where the production data of the material reached a point in its life cycle previous to the patent data reaching the same point in its life cycle. In a positive shift, the production occurs first and drives the patenting.

- iii) A negative shift is present where the production reached a point in its life cycle after the patent data reached the same point in its life cycle. Where a negative shift exists, the production is being driven by the patenting.
- c) Positive shifts may indicate innovation being driven by production and the destructive side of the innovative process while negative shifts may represent innovation driving production and the constructive side of the innovative process as proposed by Schumpeter [55]. These shifts in origin are considered as strong evidence of Stage III which verify indications of Stage III provided by strong correlation and R^2 values near one for engineering materials.
- d) A graphical representation, such as Fig. A8.5, of the relative scale, or distance, of the origin shift can be made, using a ratio of the shift and the origin, x_0 , of the production data, indicating an absolute amount that the patent or activity driving force has on the other. This ratio, called the Origin Ratio, composes the x-axis of Fig. A8.5 and is defined as $(x_0 + OS)/x_0$, where x_0 equals the production data origin and OS is the shift in origin of the modified best-fit patent data. The y-axis of Fig. 4 is the drive ratio of the material and is expressed as $(\alpha^n)_a/(\alpha^n)_p$, where $(\alpha^n)_a$ equals the activity best-fit variable alpha to the n power and $(\alpha^n)_p$ is equal to the patent best-fit variable alpha raised to the power n , in both cases n , being best-fit variables. Such a curve with the origin ratio on the x-axis and the drive ratio on the y-axis, both being non-dimensional (origin ratio is years over years and drive ratio results from $(\alpha^n)_a/(\alpha^n)_p$ which both result from equations with units of patents which cancel each other), may effectively represent innovative behavior.

- i) In Figure A8.5 the materials are divided into two groups. Group 1 materials all have drive ratios below one and origin ratios below one. Group 2 materials each have drive ratios of one or above and origin ratios of one or above.

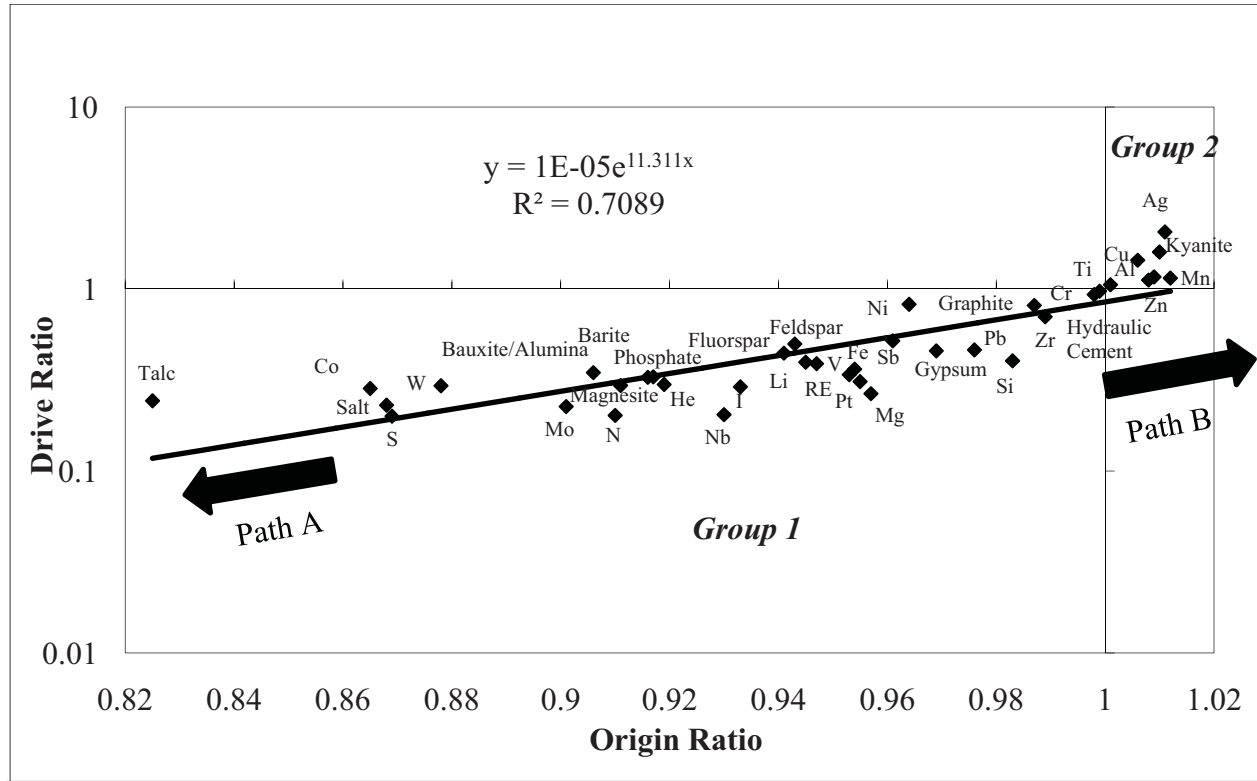


Figure A8.5. Engineering Materials Origin vs. Drive Ratio. Origin Ratio vs. Drive Ratio displaying relative strength of driving force of either patents or production activity. Note also that the cross over point occurs at 1 (y-axis). The origin shift is the shift described in section 3 between the best-fit activity and best-fit patent evaluations for each material using the common pattern equation (1). Also depicts possible paths to Stage IV.

- ii) Group 1 materials have negative origin shifts with patents driving production while Group 2 materials have positive origin shifts indicating patents are being driven by production. This may mean that Group 1 materials are in the constructive mode of innovation since patents, which are a measure of innovation, are driving the production of these materials. Group 2 may be in the destructive mode of the

- innovative process because patents, representing innovation, are being driven by production.
- iii) A point, n_0 , equal to one, near where the trend line of the data points cross the y-axis at the point where the x-axis equals one, may delineate the dividing line between the constructive and destructive modes of the innovative process.
 - iv) Such a figure may also provide an answer concerning when Stage IV is reached.
 - e) Materials such as talc, in Fig. A8.5, may leave Stage III and enter Stage IV as their origin ratios and drive ratios become farther below one as represented by the Path A arrow in Fig. 4. As both the ratios become progressively lower, patenting is strong but there is less production to drive and a point may be reached where the ratios become so small that patenting, or innovation, is no longer correlated to production and Stage IV will thus be reached. The Path B arrow in Fig. A8.5 represents materials such as silver whose origin and drive ratios are growing greater than one. In such an example, production is driving fewer and fewer patents to a point where there is little innovative activity to correlate to production. Stage IV is entered and the material becomes a commodity.
- 5) All of the preceding analyses were then applied to energy sources. It was discovered that energy sources behaved in a similar manner to engineering materials in regards to correlation, best-fit and origin shift.
- a) Energy sources displayed correlation between production (in kJ) and patents per year.
 - b) Similar life cycle and best-fit results were exhibited R^2 values were lower for production best-fit. This may be due newer energy sources, such as wind, solar and renewables, still

may only be in Stage I or II of their life cycles. The best-fit analysis and equation may be reading aspects of Stage I or II as Stage III resulting in lower R^2 values. With mature energy sources, such as oil or coal energy which have been used for many decades, the available data sets may be only displaying Stage III production leading to a lower R^2 values as well.

- c) Figure A8.6 depicts a comparison of drive ratio and origin ratio done in the same manner as Fig. A8.5 for engineering materials. The same patterns emerge as for engineering materials.
- i) Drive and origin ratios below one exist for Group 1 energy sources which indicate patents driving production inferring constructive innovative behavior.
 - ii) Group 2 is composed of energy sources with drive and origin ratios of one or greater indicating patents being driven by production and the negative mode of the innovative process.
 - iii) The dividing line between the modes of the innovative process, n_0 , appears also to be one as was the case for engineering materials.
 - iv) As with engineering materials Stage IV may be entered from either Path A from the constructive mode of innovation or via Path B from the destructive side of the innovative process becoming a commodity.
 - v) The dashed line in Fig. A8.6 represents a possible trend line for the remaining energy source data points when coal and total energy are removed or move into Stage IV.

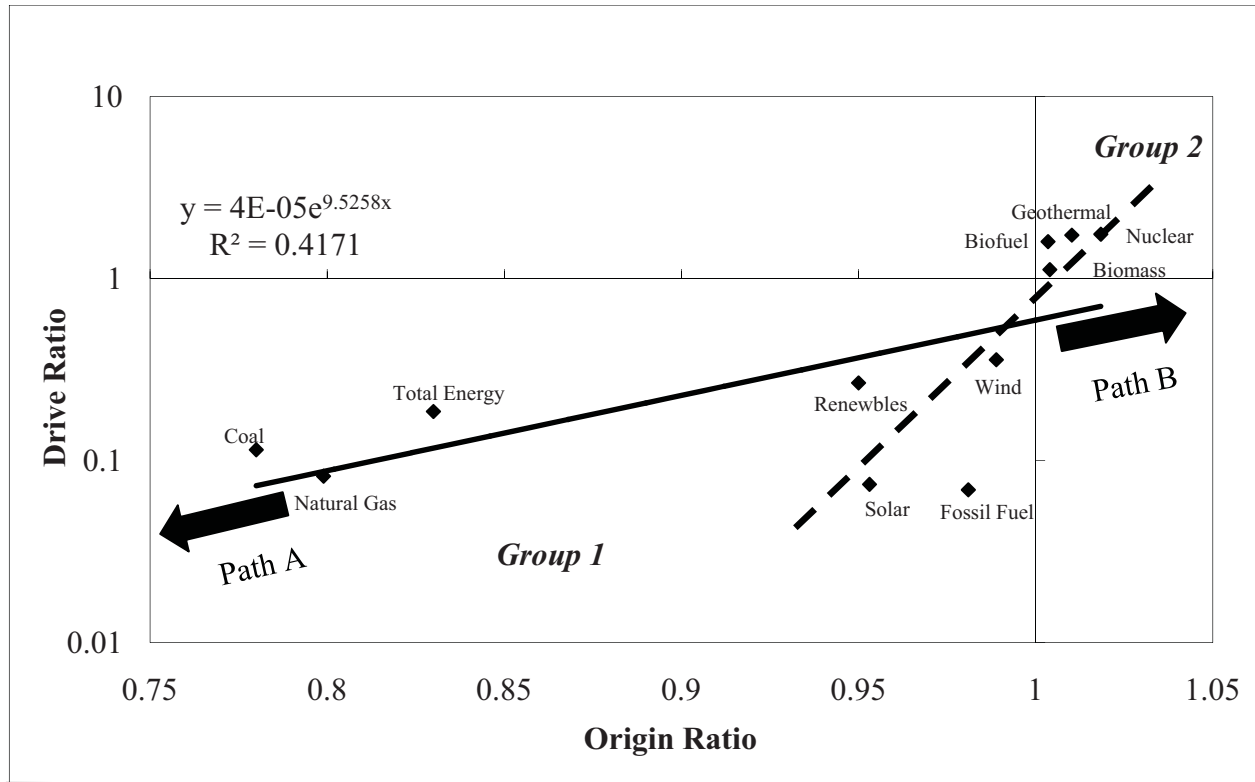


Figure A8.6. Energy Source Origin Ratio vs. Drive Origin. Displays relative strength of driving force of either patents or production activity. Possible paths A and B into Stage IV depicted by Arrows. Predicted trend line of energy source data without coal, natural gas and total energy illustrated by dashed line. R^2 for the dashed line is .6632 with an equation of $y = 1E-19e^{43.483x}$.

- 6) After evaluation of both engineering materials and energy sources it was realized that Stage III energy sources often employ or are enable by engineering and energy materials that are also in Stage III.
 - a) Table A8.1 lists Stage III energy sources and the Stage III engineering materials with their related uses in reference to the energy source.
 - b) Table A8.2 lists Stage III energy sources with the Stage III materials that are related to them. In most cases the origin shift is in the same direction and the ratios are higher or lower than one for the material as they are for the energy source. The size of the shift or ratio may not be the same but Table A8.1 is evidence of a relationship between

innovatively active energy sources being innovated by or helping to innovate Stage III materials.

Table A8.1. Examples of energy sources and the engineering materials that are innovatively active possibly due to their usage in the related energy source.

Energy source	Related Material	Usage
<i>Solar Energy</i>	Vanadium	Vanadium Redox Batteries (large power storage)
"	Silicon, Selenium	Solar Cells
<i>Wind Energy</i>	Vanadium	Vanadium Redox Batteries (large power storage)
<i>Nuclear Energy</i>	Fluorspar	Nuclear Fuel Additive
"	Uranium	Fuel
<i>Renewable Energy</i>	Graphite	Fuel Cells, Batteries
"	Nickel, Rare Earths, Cobalt	Rechargeable Batteries
"	Lithium, Cadmium, Lead	Batteries
"	Manganese	Dry Cell Batteries
"	Silver	Battery Electrodes
<i>Coal Energy</i>	Coal	Fuel
<i>Natural Gas Energy</i>	Natural Gas	Fuel

Table A8.2 Comparison of origin shifts, origin ratios and drive ratios of energy sources and the engineering and energy materials that are related to them. **Energy sources** are in kJ/year.

	Origin Shift	Origin Ratio	Drive Ratio
Solar Energy	-93 years	0.953	0.074
Vanadium	-90 years	0.954	0.363
Silicon	-30 years	0.985	0.435
Wind Energy	-22 years	0.989	0.36
Vanadium	-90 years	0.954	0.363
Nuclear energy	+36 years	1.018	1.75
Uranium	-130 years	0.94	0.61
Fluorspar	-113 years	0.941	0.444
Coal Energy	-428 years	0.780	0.115
Coal	-30 years	0.984	0.57
Natural Gas Energy	-392 years	0.799	0.082
Natural Gas	-326 years	0.832	0.22
Renewable Energy	-97 years	0.950	0.267
Graphite	-24 years	0.987	0.813
Nickel	-69 years	0.964	0.824
Rare Earths	-101 years	0.947	0.389
Cobalt	-256 years	0.865	0.284
Lithium	-106 years	0.945	0.396
Lead	-41 years	0.978	0.434
Manganese	+23 years	1.012	1.418
Silver	+21 years	1.011	2.060

7) This dissertation expands upon previous research relating to long-term life cycles in materials and extends such research into energy sources. Four-stage life cycles, which were found to

exist for the production of metals were discovered to exist in an expanded list of metals and also were shown to be present in non-metals as well as energy sources and materials. The equation used to model the life cycles of metals was modified here to render it more reliable and consistent. Similar behavior was noted for energy sources as was exhibited by engineering materials. Patterns relating to correlation, best-fit, life cycles and origin shifting were the same for both engineering materials and energy source. Shifts in the origin of the data for a modified patent best-fit revealed a possible point at which Stage III may transition into Stage IV for both materials and energy sources and verify various indicators that point towards the existence of Stage III.

The driving or driven behavior of patents, represented by an origin shift, may indicate the dual nature of innovation by identifying the constructive and destructive modes of the innovative process. A universal constant, n_0 , which is equal to one, may represent the dividing point between the two modes of the innovative process and were seen in relation to both engineering materials and energy sources.

Stage III energy sources were often found to be supported by Stage III materials that are innovatively active and seem to be contributing to the innovation of the energy source while possibly being innovated themselves at the same time. Due to lower R^2 values generated from best-fit analyses for energy sources, it is suggested here that mature energy sources such as coal and oil energy may only have production data that is in Stage III with earlier Stage I-II data being unavailable. In contrast, relatively newer energy sources, such as wind and solar energy, may still be in Stage I or II. Since the position in a life cycle can change over the years only the passage of time will reveal the exact position of these newer energy sources in their life cycles.

Advancements in the study of life cycles are presented here which may further the understanding of the innovative process in general and for engineering materials and energy sources in particular. New and original contributions are made concerning the dual nature of the innovative process and the transition from Stage III to Stage IV in long-term life cycles. It is strongly believed that the processes described here could successfully be applied to a variety of materials and systems in the future and not confined solely to the evaluation of engineering materials and energy systems.

References

- [1] Connelly, M.C. and Sekhar, J.A.; “Inventions and Innovation: A Case Study in Metals” *Key Engineering Materials* eds. Sekhar, J.A. and Dismukes J. D., 380, pp.15-39, (2008).
- [2] Yerramilli, C. and Sekhar, J.A.; “A common pattern in long-term metals production,” *Resources Policy* 31 (2006), pp. 27-36.
- [3] Sekhar J. A., & Dismukes J.; Generic innovation dynamics across the industrial technology life cycle: Platform equation modeling of invention and innovation activity, *Technological Forecasting and Social Change*, Vol. 76, Issue 1, pp 192-203, 2009.
- [4] Betz, F.; *Managing Technological Innovation: Competitive Advantage From Change*, 2nd Edition, John Wiley & Sons, New York, NY, ISBN# 0-471-22563-0, (2003).
- [5] Kondratiev, N. D.; "Die langen Wellen der Konjunktur", *Archiv fur Sozialwissenschaft und Sozialpolitik*, Vol. 56, 573-606, (1926).
- [6] Rogers, E.M.; *Diffusion of Innovations*, 1st-5th Editions, Free Press, New York, NY (1962-2003).
- [7] Lotka, A.J.; *Elements of Physical Biology*, Williams & Wilkins Company, Baltimore, Maryland, (1925).
- [8] Bass, F.M.; “A new product growth model for consumer durables,” *Management Science*, vol. 15, pp. 215-227, (1969).
- [9] Mahajan, V. and Peterson, R.A.; *Models For Innovation Diffusion*, Sage Publications, Beverly Hills, CA, (1985).
- [10] Mahajan, V., Muller, E. and Bass, F.M.; “New product diffusion models in marketing: A review and directions for research,” *Journal of Marketing*, vol. 54, pp. 1-26, (1990).
- [11] Anderson, P. and Tushman, M.L.; “Technological discontinuities and dominant designs: A cyclical model of technological change,” *Administrative Science Quarterly*, vol. 35, pp. 604-633, (1990).
- [12] Geroski, P.A.; “Models of technology diffusion,” *Research Policy*, vol. 29, pp. 603-625, (2000).
- [13] Filson, D.; “The nature and effects of technological change over the industry life cycle,” *Review of Economic Dynamics*, vol. 4, pp. 460-494, (2001).
- [14] Bass, F.M., Gordon, K., Ferguson, T.L. and Githens, M.L.; “DIRECTV: Forecasting diffusion of a new technology prior to product launch,” *Interfaces*, vol. 31, S82-S93, (2001).
- [15] Jiang, Z., Bass, F.M., and Bass, P.I.; “Virtual bass model and the left-hand data-truncation bias in diffusion of innovation studies,” *International Journal of Research in Marketing*, vol. 23, pp 93-106, (2006).
- [16] Perez, C; *Technological Revolutions and Financial Capital: The Dynamics of Bubbles and Golden Ages*, Edward Elgar Publishing, Northampton, MA, (2002).
- [17] Moore, G.A.; *Crossing the Chasm*, Harper Business Publishers, New York, NY, (1991, 1999, 2002).
- [18] Hirooka, M.; “Nonlinear dynamism of innovation and business cycles,” *J. Evol. Econ.*, vol. 13, pp 549-576, (2003).
- [19] Utterback, J.M.; “Innovation in industry and the diffusion of technology,” *Science*, vol. 183, No. 4125, Feb. 15, 1974.

- [20] Roberts, E.B.; “Managing invention and innovation,” *Research-Technology Management*, vol. 50, pp. 35-54, January – February, 2007. Note: This article was first published *Research-Technology Management*, Volume 31, pp. 11-29, 1985.
- [21] Utterback, J.M., and Fernando, F.; “Innovation, competition, and industry structure,” *Research Policy*, vol. 22, pp 1-21, 1993.
- [22] Fusfeld, H.I.; *Industry’s Future: Changing Patterns of Industrial Research*, American Chemical Society, Washington, DC, (1994).
- [23] Utterback, J.M.; *Mastering the Dynamics of Innovation*, Harvard Business School Press, Cambridge, MA, 1994.
- [24] Stokes, D.E.; *Pasteur’s Quadrant: Basic Science and Technological Innovation*, Brookings Institution Press, Washington D.C., (1997).
- [25] Bush, V.; 1946, *Science – The Endless Frontier: A Report to the President on a Program for Postwar Scientific Research*, Reviewed in SCIENCE and TECHNOLOGY POLICY YEARBOOK 1994, AAAS, Washington DC, 1994.
- [26] Myers, S. and Marquis, D.; “Successful industrial innovations,” *National Science Foundation, Report NSF 69-17*, Washington DC, (1969).
- [27] Myers, S. and Sweezy, E.E.; “Why innovations falter and fail: A study of 200 cases,” *U.S. Department of Commerce, NTIS Report PB-159-108*, (1976).
- [28] Kelly, P., and Kranzberg, P., Editors; *Technological Innovation: A Critical Review of Current Knowledge*, San Francisco Press, San Francisco, CA, 1978.
- [29] Mensch, G.; *Stalemate in Technology: Innovations Overcome The Depression*, Ballinger Publishing Company, Cambridge, MA, 1982.
- [30] Quinn, J.B.: *Innovation Explosion*, Simon & Schuster, New York, NY, (1997).
- [31] Auerswald, P.E. and Branscomb, L.M.; “Valleys of death and Darwinian Seas: financing the invention to innovation transition in the United States,” *Journal of Technology Transfer*, vol. 28, pp. 227-239, (2003).
- [32] Bromley, D.A.; “Science, technology and politics,” *Technology in Society*, vol. 24, pp. 9-26, (2002).
- [33] Bromley, D.A.; “Technology policy”, *Technology in Society*, vol. 26, pp. 455-468, (2004).
- [34] Godin, B.; “The linear model of innovation: The historical construction of an analytical framework,” *Science, Technology & Human Values*, vol. 31, Number 6, pp. 639-667, November (2006).
- [35] Grove, A.S.; *Only The Paranoid Survive: How To Exploit the Crisis Points that Challenge Every Company*, Doubleday, New York, NY, (1996).
- [36] Chesbrough, H. and Rosenbloom, R.S.; “The role of the business model in capturing value from innovation: Evidence from Xerox Corporation’s technology spin-off companies,” *Industrial and Corporate Change*, vol. 11, No. 3, pp. 529-555, (2002).
- [37] Chesbrough, H.W.; *Open Innovation: The New Imperative for Creating and Profiting from Technology*, Harvard Business School Press, Boston, MA (2003).
- [38] Chesbrough, H. and Spohrer, J.; “A research manifesto for services science,” *Communications of the ACM*, vol. 49, No. 7, pp. 33-40, July (2006).
- [39] Burt, R.S.; *The Network Structure of Social Capital*, in *Research in Organizational Behavior*, R. I. Sutton, B. M. Staw, Editors, JAI Press, Greenwich, CT, pp. 345-423, (2000).
- [40] Burt, R.S.; “Bridge decay”, *Social Networks*, vol. 24, issue 4, pp. 333-363, (2002).

- [41] Smits, R.; "Innovation studies in the 21st century: Questions from a user's perspective," *Technological Forecasting and Social Change*, vol. 69, pp. 861-883, (2002).
- [42] Tellis, G.J.; "Disruptive technology or visionary leadership," *Journal of Product Innovation Management*, vol. 23, issue 1, 34-38, (2006).
- [43] Gerybadze, A. and Reger, G.; "Globalization of R&D: Recent changes in the management of innovation in transnational corporation," *Research Policy*, vol. 28, pp. 251-274, (1999).
- [44] Forrest, J.E.; "Models of the process of technological innovation," *Technology Analysis and Strategic Management*, vol. 4, no. 4, 439-452, (1991).
- [45] Rothwell, R.; "Towards the fifth-generation innovation process," *International Marketing Review*, vol. 11, pp. 7-31, 1994.
- [46] Porter, M.E.; *The Competitive Advantage of Nations*, The Free Press, New York, New York, 1990.
- [47] Leifer, R., McDermot, C.M., O'Connor, G.C., Peters, L.S., Rice, M., and Veryzer, R.W.; *Radical Innovation: How Mature Companies Can Outsmart Upstarts*, Harvard Business School Press, Boston, MA, (2000).
- [48] Christensen, C.M.; "The ongoing process of building a theory of disruption," *Journal of Product Innovation Management*, vol. 23, issue 1, 39-55, (2006).
- [49] Drucker, P.E.; "The discipline of Innovation," *Harvard Business Review*, pp. 67-71, May-June 1985.
- [50] Voss, C.A.; "The need for a field of study of implementation of innovations," *J. Prod. Innov. Management*, vol. 4, pp. 266-271, (1985).
- [51] Age, J.O.; "Development of a model for technological innovation process," *Technology Management*, vol. 2, 291-292, (1995).
- [52] Rich, B.R. and Janos, L.; *Skunk Works: A Personal Memoir of My Years at Lockheed*, Little, Brown and Company, New York, NY, (1994).
- [53] Friedman, T.L.; *The World is Flat: A Brief History of the Twenty-First Century*, Farrar, Straus and Giroux, New York, NY, (2005).
- [54] "Innovate America National Innovation Initiative Report", *Council on Competitiveness*, Washington, DC, 1st Edition, Dec. 2004; and 2nd Edition, <http://www.compete.org/>, (2005).
- [55] McGraw, T.K.; *Prophet of Innovation: Joseph Schumpeter and Creative Destruction*. The Belknap Press of Harvard University Press, Cambridge Mass, (2007).
- [56] Schumpeter, J.A.; *Capitalism, Socialism and Democracy*. Harper & Brothers Publishers, New York, (1942).
- [57] Erwin, D.H. and Krakauer, D.C.; "Insights into innovation," *SCIENCE* 304, pp. 1117-1119, (2004).
- [58] Brandt, A.; "Patent overload hampers tech innovation," *WWW.PCWORLD.COM*, pg. 24 April, 2006.
- [59] Schneiderman, R.; "Patents cuffing innovation? Patent claims are threatening what have been accepted as royalty-free standards," *Electronic Design* 53 (9), (2005).
- [60] Reguly, E.; "Patent protection a threat to innovation," *The Globe and Mail*, January 1, 2006.
- [61] Branstetter, L.; "Do stronger patents induce more local innovation?," *Journal of International Economic Law* 7 (2) pp. 359-370, (2004).

- [62] Sakakibara, M. and Branstetter, L.; “Do stronger patents induce more innovation? Evidence from the 1988 Japanese patent law reforms,” *RAND Journal of Economics* 32 (1), pp.77-100, (2001).
- [63] Acs, Z.J. and Audretsch, D.B.; “Patents as a measure of innovative activity,” *Kyklos* 42 (2),pp. 171-180, (1989).
- [64] Archibugi, D. and Pianta, M.; “Measuring technological change through patents and innovation surveys,” *Technovation* 16 (9), pp. 451-468, (1996).
- [65] Rogers, M.; “The definition and measurement of innovation,” *Melbourne Institute of Applied Economic and Social Research, The University of Melbourne, Melbourne Institute Working Paper No. 10/98*, (1998).
- [66] Alegre, J., Lapiedra, R. and Chiva, R.; “A measurement scale for product innovation performance,” *European Journal of Innovation Management* 9 (4), pp. 333-346, (2006).
- [67] Van Der Panne, G.; “Issues in measuring innovation,” *Scientometrics* 71 (3), pp. 495-507, (2007).
- [68] Walker, R.M., Jeanes, E. and Rowlands, R.; “Measuring innovation – applying the literature – based innovation output indicator to public services,” *Public Administration* 80 (1), pp. 201-214, (2002).
- [69] Blankley, W. and D. Kaplan: Innovation and South African Industry: What are We Trying to Measure? *South African Journal of Science* 94 (2), pp. 50-53, (1998).
- [70] Sherry, E.F. and Teece, D.J.; “Royalties, evolving patent rights, and the value of innovation,” *Research Policy* 33, pp.179-191, (2003).
- [71] Wilson, A.L., Ramamurthy, K. and Nystrom, P.C.; “A multi-attribute measurement for innovation adoption: The context of imaging technology,” *IEEE Transactions on Engineering Management* 46 (3), pp. 311-320, (1999).
- [72] Green, S.G., Gavin, M.B. and Aiman-Smith, L.; “Assessing a multidimensional measure of radical technological innovation,” *IEEE Transactions on Engineering Management* 42(3), pp. 203-214, (1995).
- [73] Worgan, A. and Nunn, S.; “Exploring a complicated labyrinth: Some tips on using patent data to measure urban and regional innovation,” *Economic Development Quarterly* 16 (3), pp. 229-236, (2002).
- [74] Wu, Y.J. and Lee, P.; “The use of patent analysis in assessing ITS innovations: US, Europe and Japan,” *Transportation Research part A* 41, pp. 568-586, (2006).
- [75] McAleer, M. and Slottje, D.; “A new measure of innovation: The patent success ratio,” *Scientometrics* 63 (3), pp. 421-429, (2005).
- [76] Jaffe, A.B., Fogarty, M.S. and Banks, B.A.; “Evidence from patents and patent citations on the impact of NASA and other federal labs on commercial innovation,” *The Journal of Industrial Economics* 46 (2), pp. 183-204, (1998).
- [77] Alcacer, J. and Gittleman, M.; “Patent citations as a measure of knowledge flows: The influence of examiner citations,” *The Review of Economics and Statistics* 88 (4), pp. 774-779, (2006).
- [78] United States Geologic Survey. usgs.gov.
- [79] United States Geologic Survey. <http://minerals.usg.gov/minerals>.
- [80] European Patent Office. esp@cenet.
- [81] Walpole, R.E., Myers, R.H. and Myers, S; *Probability and Statistics for Engineers and Scientists*, Sixth Edition. Prentice-Hall Inc., NJ (1998).

- [82] Miller, I., Freund, J.E. and Johnson, R.A.; *Probability and Statistics for Engineers*, Fourth Edition. Prentice-Hall, Inc., NJ (1990).
- [83] Connelly, M.C.; “The Relationship Between Patents and Technical Innovation: Innovation measurement as applied to metals.” MS Thesis, University of Cincinnati, 2007.
- [84] Connelly, M.C., Dismukes, J.P. and Sekhar, J.A.; “New Relationships Between Patents and Technological Innovation: Modeling patent Activity as a Driver of Innovation.” PICMET ‘09 Proceedings, Portland, Oregon, August 2009. For shorter product life cycles See also [Levitt, T.: “Exploit the product life cycle,” *Harvard Business Review*, November-December (1965)].
- [85] Freeman. C., (Ed.): *Long Wave Theory*. Elgar Publishing Limited, Cheltenham, U.K (1996).
- [86] Jenner, R. A.; “Real wages, business cycles and new production patterns,” *Small Business Economics* 23, 441-452 (2004).
- [87] Keklik, M.: *Schumpeter Innovation and Growth: Long Cycle Dynamics in Post WWII American Manufacturing Industries*. Ashgate, Vermont (2003).
- [88] Phillips, K. L. and Wrase, J.; “Is Schumpeterian ‘creative destruction’ a plausible source of endogenous real business cycle shocks?,” *Journal of economic Dynamics & Control* 30, 1885-1913 (2006).
- [89] Wong, H. and Ellis, P. D.; “Is market orientation affected by the product life cycle?,” *Journal of World Business* 42, 145-156 (2007).
- [90] United States Geologic Survey. usgs.gov.; *Minerals Handbook*, (2007).
- [91] U.S. Energy Information Administration [EIA]. <http://www.eia.doe.gov>.
a) For natural gas : http://tonto.eia.doe.gov/dnav/ng/ng_prod_sum_dc_u_NUS_a.htm
b) For oil: http://tonto.eia.doe.gov/dnav/pet/pet_crd_crpdn_adc_mbbl_a.htm
c) For renewable energy sources biomass, solar, hydroelectric, geothermal, wind, biofuel, wood and total renewables: <http://www.eia.doe.gov/emeu/aer/renew.html>
d) For uranium: <http://www.eia.doe.gov/emeu/aer/nuclear.html>
e) For fossil fuel and nuclear: <http://www.eia.doe.gov/emeu/aer/overview.html>
f) For coal: <http://www.eia.doe.gov/emeu/aer/coal.html> and http://www.eia.doe.gov/cneaf/coal/page/coal_production_review.pdf
- [92] Berry, D.; “Innovation and the price of wind energy in the US,” *Energy Policy* 37, pp. 4493-4499, (2009).
- [93] Dismukes, J.P., Miller, L.K. and Bers, J.A.; “The industrial life cycle of wind energy electrical power generation ARI methodology of life cycle dynamics,” *Technological Forecasting & Social Change* 76, pp. 178-191, (2009).
- [94] Inoue, Y. and Miyazaki, K.; “Technological innovation and diffusion of wind power in Japan,” *Technological Forecasting & Social Change* 75, pp. 1303-1323, (2008).
- [95] Muylaert de Araújo, M.S. and Vasconcelos de Freitas, M.A.; “Acceptance of renewable energy innovation in Brazil-case study of wind energy,” *Renewable and Sustainable Energy Reviews* 12, pp. 548-591, (2008).
- [96] Shikha, Bhatti, T.S. and Kothari, D.P.; “New Horizons for Offshore Wind Energy: Shifting Paradigms and Challenges,” *Energy Sources* 27, pp. 349-360, (2005).
- [97] Kobos, P.H., Erickson, J.D. and Drennen, T.E.; “Technological learning and renewable energy costs: implications for US renewable energy policy,” *Energy Policy* 34, pp. 1654-1658, (2006).

- [98] Harborne, P. and Hendry, C.; "Pathways to commercial wind power in the US, Europe and Japan: The role of demonstration projects and field trials in the innovation process," *Energy Policy* 37, pp. 3580-3595, (2009).
- [99] Wüstenhagen, R., Wolsink, M. and Bürer, M.J.; "Social acceptance of renewable energy innovation: An introduction to the concept," *Energy Policy* 35, pp. 2683-2691, (2007).
- [100] Negro, S.O. and Hekkert, M.P.; "Explaining the success of emerging technologies by innovation system functioning: the case of biomass digestion in Germany," *Technology Analysis & Strategic Management* 20, pp. 465-482, (2008).
- [101] Louime, C. and Uckelmann, H.; "Cellulosic Ethanol: Securing the Planet Future Energy Needs," *International Journal of Molecular Sciences* 9, pp. 838-841, (2008).
- [102] Kelly-Yong, T.L., Lee, K.T., Mohamed, A.R. and Bhatia, S.; "Potential of hydrogen from oil palm biomass as a source of renewable energy worldwide," *Energy Policy* 35, pp. 5692-5701, (2007).
- [103] van der Laak, W.W.M., Raaven, R.P.J.M. and Verbong, G.P.J.; "Strategic niche management for biofuels: Analyzing past experiments for developing new biofuel policies," *Energy Policy* 35, pp. 3213-3225, (2007).
- [104] Wonglimpiyarat, J.; "Technological change of the energy innovation system: From oil-based to bio-based energy," *Applied Energy* 87, pp. 749-755, (2010).
- [105] Vertès, A.A., Inui, M. and Yukawa, H.; "Technological Options for Biological Fuel Ethanol," *Journal of Molecular Microbiology and Biotechnology* 15, pp. 16-30, (2008).
- [106] Trancik, J.E.; "Scale and innovation in the energy sector: a focus on photovoltaics and nuclear fission," *Environmental Research Letters* 1, 014009 (7 pp.), (2006).
- [107] Lee, T.J., Lee, K.H. and Oh, K.B.; "Strategic environments for nuclear energy innovation in the next half century," *Progress in Nuclear Energy* 49, pp. 397-408, (2007).
- [108] Li, J.; "Scaling up concentrating solar thermal technology in China," *Renewable and Sustainable Energy Reviews* 13, pp. 2051-2060, (2009).
- [109] Faiers, A., Neame, C. and Cook, M.; "The adoption of domestic solar-power systems: Do consumers assess product attributes in a stepwise process?" *Energy Policy* 35, pp. 3418-3423, (2007).
- [110] Faiers, A. and Neame, C.; "Consumer attitudes towards domestic solar power systems," *Energy Policy* 34, pp. 1797-1806, (2006).
- [111] Huang, A.Y.J. and Liu, R.H.; "Learning for supply as a motive to be the early adopter of a new energy technology: A study on the adoption of stationary fuel cells," *Energy Policy* 36, pp. 2143-2153, (2008).
- [112] Suurs, R.A.A., Hekkert, M.P. and Smits, R.E.H.M.; "Understanding the build-up of a technological innovation system around hydrogen and fuel cell technologies," *International Journal of Hydrogen Energy* 34, pp. 9639-9654, (2009).
- [113] Hellman, H.L. and van der Hoed, R.; "Characterising fuel cell technology: Challenges of the commercialisation process," *International Journal of Hydrogen Energy* 32, pp. 305-315, (2007).
- [114] Rourke, F.O., Boyle, F. and Reynolds, A.; "Tidal energy update 2009," *Applied Energy* 87, pp. 398-409, (2010).
- [115] Bañales-López, S. and Norberg-Bohm, V.; "Public policy for energy technology innovation: A historical analysis of fluidized bed combustion development in the USA," *Energy Policy* 30, pp. 1173-1180, (2002).

- [116] Tsoutsos, T.D. and Stamboulis, Y.A.; “The sustainable diffusion of renewable energy technologies as an example of an innovation-focused policy,” *Technovation* 25, pp. 753-761, (2005).
- [117] de Vries, B.J.M, van Vuuren, D.P. and Hoogwijk, M.M.; “Renewable energy sources: Their global potential for the first-half of the 21st century at a global level: An integrated approach,” *Energy Policy* 35, pp. 2590-2610, (2007).
- [118] Bergek, A., Jacobsson, S. and Sandén, B.A.; “‘Legitimation’ and ‘development of positive externalities’: two key processes in the formation phase of technological innovation systems,” *Technology Analysis & Strategic Management* 20, No. 5, pp. 575-592, (2008).
- [119] Sovacool, B.K.; “Resolving the impasse in American energy policy: The case for a transformational R&D strategy at the U.S. Department of energy,” *Renewable and Sustainable Energy Reviews* 13, pp. 346-361, (2009).
- [120] Bonilla, S.H., Almeida, C.M.V.B., Giannetti, B.F. and Huisingh, D.; “The roles of cleaner production in the sustainable development of modern societies: an introduction to this special issue,” *Journal of Cleaner Production* 18, pp. 1-5, (2010).
- [121] Schmidt, R.C. and Marschinski, R.; “A model of technological breakthrough in the renewable energy sector,” *Ecological Economics* 69, pp. 435-444, (2009).
- [122] Norberg-Bohm, V.; “Creating Incentives for Environmentally Enhancing Technological Change: Lessons From 30 Years of U.S. Energy Technology Policy,” *Technological Forecasting and Social Change* 65, pp. 125-148, (2000).
- [123] Narayanamurti, V., Anadon, L.D. and Sagar, A.D.; “Transforming Energy Innovation,” *Issues in Science & Technology*, pp. 57-64, (Fall 2009).
- [124] Bürer, M.J. and Wustenhagen, R.; “Which renewable energy policy is a venture capitalist’s best friend? Empirical evidence from a survey of international cleantech investors,” *Energy Policy* 37, pp. 4997-5006, (2009).
- [125] Bonvillian, W.B. and Weiss, C.; “Stimulating Innovation in Energy Technology,” *Issues in Science & Technology*, pp. 51-56, (Fall 2009).
- [126] Wang, T.J. and Liu, S.Y.; “Shaping and exploiting technological opportunities: The case of technology in Taiwan,” *Renewable Energy* 35, pp. 360-367, (2010).
- [127] Connelly, M.C. and Sekhar, J.A.; “A Case Study in Metals for Inventions and Innovations” PICMET ‘08 Proceedings, Cape Town, South Africa, (July 2008).
- [128] Isaacs, A., Daintith and Martin, E., Eds.; *Dictionary of Science*, Grange Books, London, (2005) Originally published as *Oxford Dictionary of Science*, Market House Books Ltd, (2003).