

The Longevity of Famous People from Hammurabi to Einstein Online Appendix

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Abstract

We build a new sample of 300,000 famous people born between Hammurabi's epoch and Einstein's cohort, including their vital dates, occupations, and locations from the Index Bio-bibliographicus Notorum Hominum. We discuss and control for selection and composition biases. We show using this long-running consistent database that there was no trend in mortality during most of human history, confirming the existence of a Malthusian epoch; we date the beginning of the steady improvements in longevity to the cohort born in 1640-9, clearly preceding the Industrial Revolution, lending credence to the hypothesis that human capital may have played a significant role in the take-off to modern growth; we find that this timing of improvements in longevity concerns most countries in Europe and most skilled occupations.

JEL Classification Numbers: J11, I12, N30, I20, J24.

Keywords: Longevity, Notoriety, Malthus, Elite, Compensation Effect of Mortality, Enlightenment, Europe.

A Lifespan Precision

To measure the quality of the individual lifespan data, in this section we show two different statistics: the frequency of observations with imprecise vital dates and the heaping index.

The IBN adds the indications “c.”, for *circa*, or “?” to the vital dates when the years of birth or death are not known with certainty. It may also be that more than one date is reported. We retained all the imprecise observations (taking the mean if there was more than one date), but created a discrete variable called *precision*, allocating a value of one when the lifespan was imprecise, zero otherwise. Figure A.1 shows the fraction of imprecise observations by decade. Individual lifespans measured by the IBN were highly imprecise until the end of the Middle Ages; the degree of imprecision then moves to zero as the sample reaches the 19th century.

When vital data are not known with certainty, biographers (or concerned persons themselves) often approximate them by rounding the year of death or birth to a number finishing in 0 or 5. Moreover, in the particular case of famous people, for obvious reasons, years of birth are likely to be more uncertain than years of death. The heaping index measures the frequency of observations with vital dates finishing in 0 or 5; it is commonly normalized by multiplying the frequency by 5. A heaping index close to unity shows that the vital data are very precise. Figure A.2 shows birth and death heaping indexes by decades up to 1879.¹ The death date heaping index is low, indicating that the dates of death of famous people were well known. Birth dates were much more uncertain, as the heaping index is about 2.5 before 1450, indicating that there are 2.5 times more dates finishing in 0 or 5 than there should be. Improvements in the birth year heaping index seem to start around 1450. This observation is consistent with the findings of De Moor and Zuijderduijn (2013) that numeracy levels among the well-to-do in the early modern period were very low (in the Netherlands). By 1700, the gap between birth and death heaping has decreased and both indexes fluctuate around one. If, following A’Hearn, Baten, and Crayen (2009), we interpret the age heaping index as a measure of human capital (consistently with the robust correlation between age heaping and literacy at both the individual and aggregate level), our findings support the hypothesis that there was a major increase in human capital preceding the industrial revolution.

¹Notice that heaping has no sense before 800, when the dating system starting at the birth of Jesus of Nazareth became widely used.

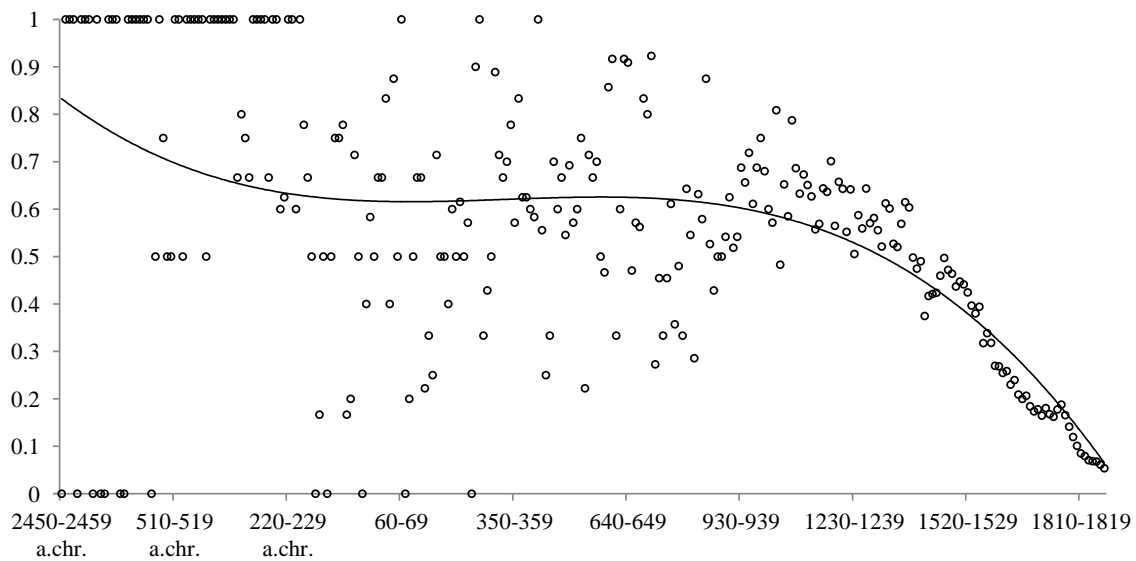


Figure A.1: Frequency of Imprecise Observations

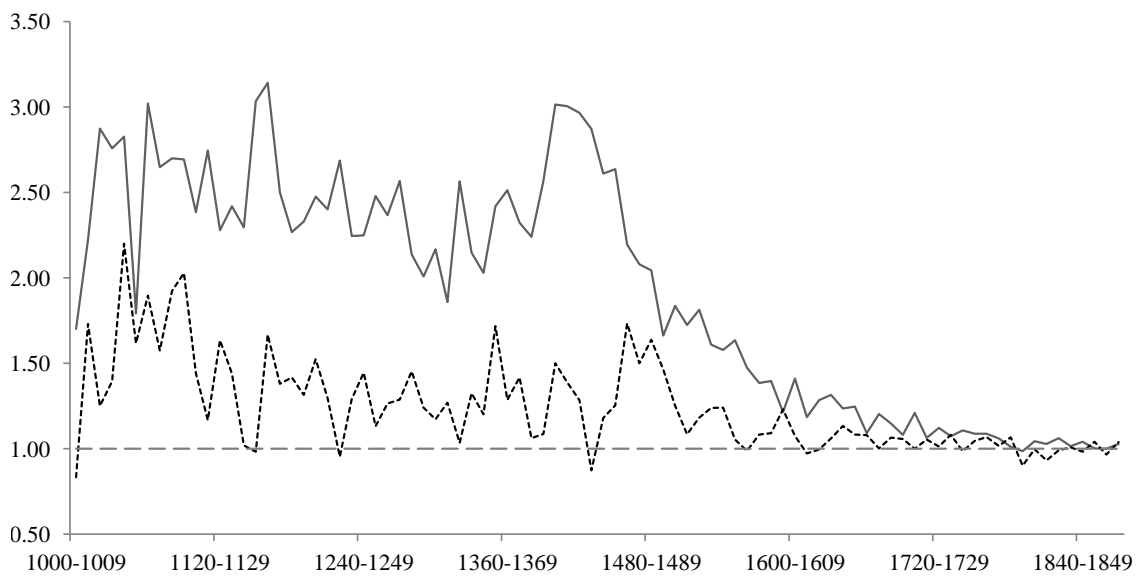


Figure A.2: Heaping Index. Birth year (solid line), death year (dashed line)

B Longevity vs Life Expectancy

The population of currently famous people alive at the beginning to time s , N_s , evolves according to:

$$N_{s+1} = N_s - d_s + I_{s+1} \quad (1)$$

where d_s is the number of deaths between age s and age $s+1$, and I_{s+1} is the number of new people gaining celebrity over the same age interval. When computing life expectancy, N_s is the population at risk. Unfortunately, we do not observe N_s since we do not know when people become famous (except in special cases for which a nomination is required) and I_{s+1} is unobserved.

The life expectancy at age a of this population is:

$$E_a = \sum_{s=a}^T (s-a)m_{s,a}$$

where T is the maximum number of periods one can live, and $m_{s,a}$ is the true probability of dying at age s conditionally on being alive at age a :

$$m_{s,a} = \frac{d_s}{N_s} \times S_{s,a}.$$

$S_{s,a}$ is the probability of reaching age s if one has reached age a . It follows:

$$S_{s+1,a} = S_{s,a} \times \left(1 - \frac{d_s}{N_s}\right) = S_{a,a} \times \prod_{j=a}^s \left(1 - \frac{d_j}{N_j}\right) = \prod_{j=a}^s \left(1 - \frac{d_j}{N_j}\right)$$

Notice that, contrary to the case where all individual belong to the population at age a and can be followed until death, $m_{s,a} \neq d_s/N_a$.

One can rewrite the population at risk as:

$$N_{s+1} = S_{s+1,a} \left(N_a + \sum_{j=a+1}^{s+1} \frac{I_j}{S_{j,a}} \right). \quad (2)$$

We denote the population of all famous people aged s by \hat{N}_s . This population includes everyone that is or will become famous. Contrary to N_s , we observe \hat{N}_s . Its dynamics are given by

$$\hat{N}_{s+1} = \hat{N}_s - d_s \quad (3)$$

Using equations (1) and (3), we can compute the gap between total population \hat{N} and population at risk N as

$$\hat{N}_s - N_s = \hat{N}_{s+1} - N_{s+1} + I_{s+1}$$

Iterating forward, we get

$$\hat{N}_s - N_s = \hat{N}_T - N_T + \sum_{j=s+1}^T I_j = \sum_{j=s+1}^T I_j \quad (4)$$

where T is the date at which all famous people have been discovered, i.e., $I_j = 0 \forall t > T$. It implies $\hat{N}_T = N_T$. The above equation reflects the idea that \hat{N}_s incorporates all the people that are not yet famous but will be. We can define $\hat{m}_{s,a}$ as the observable probability of dying at age s conditionally on being alive at age a :

$$\hat{m}_{s,a} = \frac{d_s}{\hat{N}_a}.$$

We have the following property:

$$\sum_{s=a}^T d_s = \hat{N}_a, \quad \sum_{s=a}^T \hat{m}_{s,a} = 1.$$

We can measure the mean lifetime conditionally on being alive at a :

$$L_a = \sum_{s=a}^T s \hat{m}_{s,a}$$

which we call “longevity”.

There is a gap G_a between the expected length of life $a + E_a$:

$$G_a = L_a - E_a - a = \sum_{s=a}^T (s - a)(\hat{m}_{s,a} - m_{s,a}).$$

This gap comes from the fact that we cannot compute the correct mortality rates $m_{s,a}$ as we do not know the population at risk. Replacing $\hat{m}_{s,a}$ by its value, we obtain

$$G_a = \sum_{s=a}^T (s - a) \left(\frac{d_s}{N_s} \frac{N_s}{\hat{N}_a} - m_{s,a} \right) = \sum_{s=a}^T (s - a) m_{s,a} \left(\frac{N_s}{\hat{N}_a S_{s,a}} - 1 \right)$$

Using (2) we get:

$$\begin{aligned} G_a &= \sum_{s=a}^T (s-a) m_{s,a} \left(\frac{S_{s,a} \left(N_a + \sum_{j=a+1}^s \frac{I_j}{S_{j,a}} \right)}{\hat{N}_a S_{s,a}} - 1 \right) \\ &= \sum_{s=a}^T (s-a) m_{s,a} \left(\frac{N_a + \sum_{j=a+1}^s \frac{I_j}{S_{j,a}}}{\hat{N}_a} - 1 \right) \end{aligned}$$

Using (4) we get:

$$\begin{aligned} G_a &= \sum_{s=a}^T (s-a) m_{s,a} \left(\frac{\hat{N}_a - \sum_{j=a+1}^T I_j + \sum_{j=a+1}^s \frac{I_j}{S_{j,a}}}{\hat{N}_a} - 1 \right) \\ &= \sum_{s=a}^T (s-a) m_{s,a} \left(\sum_{j=a+1}^s \frac{I_j}{\hat{N}_a S_{j,a}} - \sum_{j=a+1}^T \frac{I_j}{\hat{N}_a} \right). \end{aligned}$$

Here is how the gap depends on the process leading to notoriety $\{I_j\}_{j=a..T}$:

$$G_a = \sum_{s=a}^T (s-a) m_{s,a} \left(\sum_{j=a+1}^s \frac{I_j(1-S_{j,a})}{\hat{N}_a S_{j,a}} - \sum_{j=s+1}^T \frac{I_j}{\hat{N}_a} \right).$$

To illustrate the effect of a change in the age at which people become famous, suppose that a proportion μ of all famous people are already famous at age a and that the proportion $1 - \mu$ gets famous at age $f > a$. Then the bias is:

$$B_a = \sum_{s=a}^T (s-a) m_{s,a} \left(\frac{(1-\mu)\hat{N}_a}{\hat{N}_a S_{f,a}} \right) = \frac{1-\mu}{S_{f,a}} E_a.$$

The bias is therefore proportional to life expectancy, with the proportionality factor increasing in f (as $S_{f,a}$ is decreasing in f) and decreasing in μ . If age at notoriety f changes, we have:

$$\frac{\partial B_a}{\partial f} = -\frac{1-\mu}{S_{f,a}^2} E_a \frac{\partial S_{f,a}}{\partial f}.$$

The derivatives depends on the slope of the survival function at age f . If it is not too decreasing at f ($\frac{\partial S_{f,a}}{\partial f}$ is small), for example when S is concave and f is low enough, the effect on the bias will be small.

C Occupation categories

Arts and métiers: actor, artist, cantor, collector, composer, designer, dramatist, engraver, goldsmith, illustrator, kapellmeister, lithograph, musician, organist, painter, pewterer, pianist, poet, regisseur, sculptor, singer, violinmaker and violinist.

Business: antiquary, bookseller, banker, printer, publicist, businessman, director, editor, farmer, librarian, industrialist, merchant, trader, manufacturer and wholesaler.

Education: author, academician, dean, lecturer, professor, rector, scholar, student, teacher and writer.

Humanities: archaeologist, classicist, economist, historian, journalist, orientalist, pedagogue, philologist, philosopher and translator.

Law and government: administrator, adviser, ambassador, bailiff, beamter, congressman, consul, councillor, deputy, diplomat, governor, inspector, judge, jurist, lawyer, magistrato, mayor, minister, notary, politician, prefect, president, procureur, secretary, senator and sheriff.

Military: admiral, brigadier-general, captain, colonel, commander, fighter, general, lieutenant, lieutenant-colonel, major, major-general, marshal, military, officer and soldier.

Nobility: baron, chamberlain, duke, earl, king, knight, lord, noble, prince and queen.

Religious: abbot, archbishop, archdeacon, capuchin, cardinal, clergyman, deacon, franciscan, jesuit, martyr, missionary, pastor, piarist, preacher, priest, rabbi, theologian and vicar.

Sciences: agronomist, architect, astronomer, botanist, builder, cartographer, chemist, doctor, engineer, geographer, geologist, inventor, mathematician, naturalist, pharmacist, physician, physicist, surgeon and zoologist.

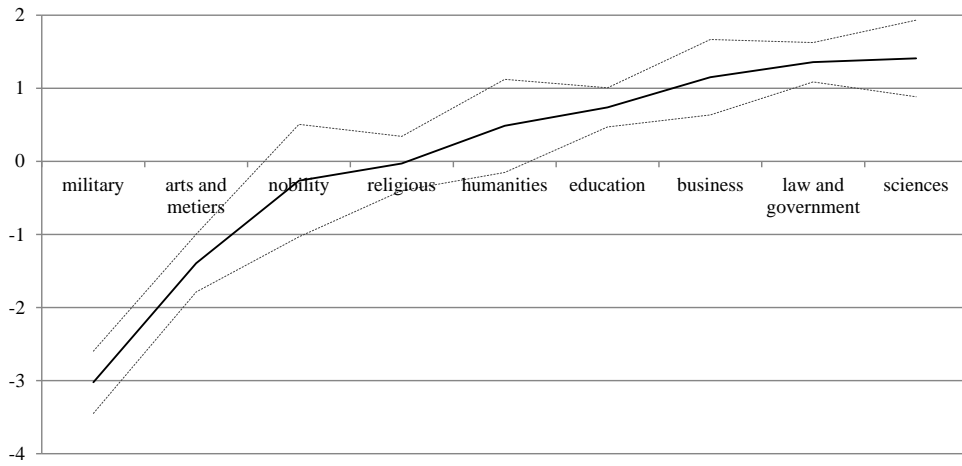


Figure A.3: Conditional Longevity. Main occupational groups

D Analysis of the Residuals

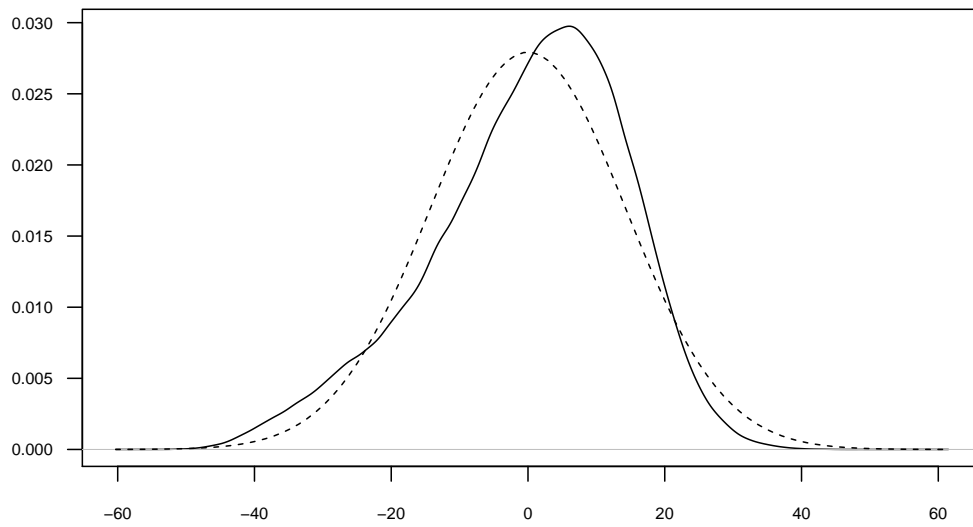


Figure A.4: Kernel Density of the Residuals (solid) and Normal density (dashes)

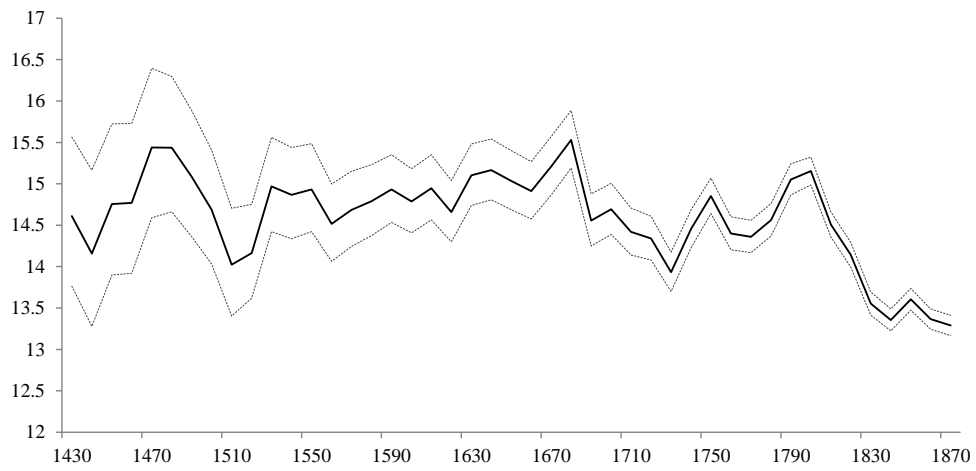


Figure A.5: Standard Deviation of Residuals by Decade, and 95% confidence interval

E Longevity per Year of Death

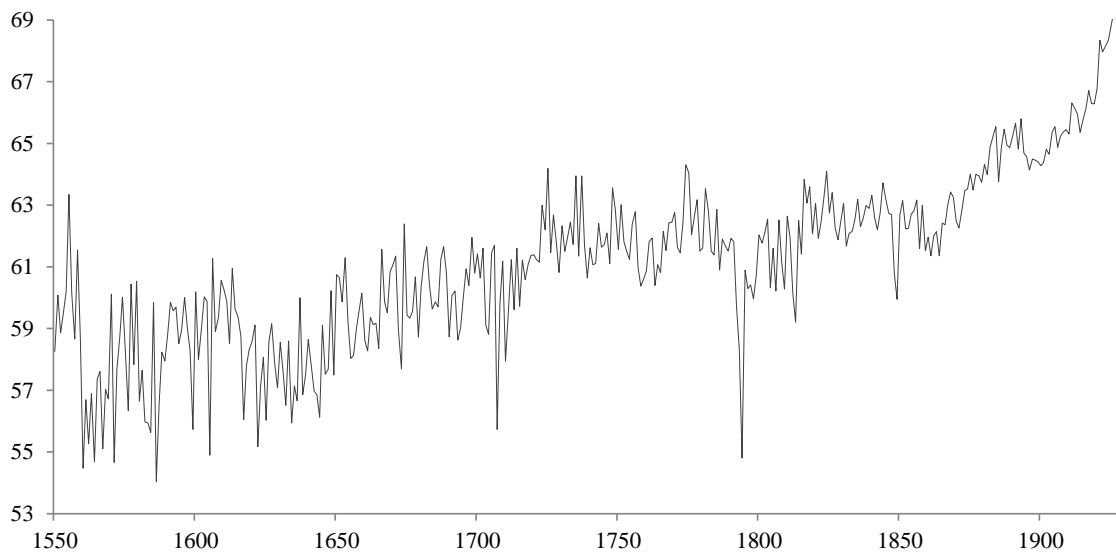


Figure A.6: Longevity per Year of Death

F 1600 cohorts

cohort #	years	# obs
1	2450 BCE - 1040 CE	1,611
2	1041 - 1254	1,602
3	1255 - 1360	1,634
4	1361 - 1415	1,617
5	1416 - 1450	1,737
6	1451 - 1481	1,619
7	1482 - 1502	1,600
8	1503 - 1520	1,676
9	1521 - 1534	1,636
10	1535 - 1546	1,675
11	1547 - 1557	1,641
12	1558 - 1566	1,627
13	1567 - 1575	1,765
14	1576 - 1583	1,708
15	1584 - 1590	1,660
16	1591 - 1597	1,723
17	1598 - 1603	1,773
18	1604 - 1610	1,926
19	1611 - 1616	1,616
20	1617 - 1622	1,740
21	1623 - 1628	1,743
22	1629 - 1633	1,621
23	1634 - 1639	1,870
24	1640 - 1644	1,660
25	1645 - 1649	1,621
26	1650 - 1654	1,731
27	1655 - 1659	1,663
28	1660 - 1664	1,862
29	1665 - 1669	1,710
30	1670 - 1674	1,846
31	1675 - 1679	1,730

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cohort #	years	# obs
32	1680 - 1683	1,609
33	1684 - 1688	1,888
34	1689 - 1693	1,903
35	1694 - 1697	1,713
36	1698 - 1701	1,872
37	1702 - 1705	1,641
38	1706 - 1709	1,680
39	1710 - 1713	1,946
40	1714 - 1717	2,023
41	1718 - 1720	1,662
42	1721 - 1724	2,106
43	1725 - 1727	1,763
44	1728 - 1730	1,879
45	1731 - 1733	1,892
46	1734 - 1736	2,046
47	1737 - 1739	1,994
48	1740 - 1742	2,208
49	1743 - 1745	2,298
50	1746 - 1748	2,284
51	1749 - 1750	1,874
52	1751 - 1752	1,795
53	1753 - 1754	1,793
54	1755 - 1756	1,901
55	1757 - 1758	1,834
56	1759 - 1760	1,907
57	1761 - 1762	1,840
58	1763 - 1764	2,042
59	1765 - 1766	2,125
60	1767 - 1768	1,978
61	1769 - 1770	2,231
62	1771 - 1772	2,054
63	1773 - 1774	2,074

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cohort #	years	# obs
64	1775 - 1776	2,041
65	1777 - 1778	2,060
66	1779 - 1781	2,039
67	1781 - 1782	1,867
68	1783 - 1784	1,997
69	1785 - 1786	2,146
70	1787 - 1788	2,255
71	1789 - 1790	2,390
72	1791 - 1792	2,240
73	1793 - 1794	2,436
74	1795 - 1796	2,485
75	1797 - 1798	2,620
76	1799 - 1800	2,950
77	1801 - 1802	3,061
78	1803 - 1804	3,017
79	1805 - 1806	3,152
80	1807 - 1808	3,161
81	1809 - 1810	3,322
82	1811	1,688
83	1812	1,743
84	1813	1,611
85	1814	1,642
86	1815	1,795
87	1816	1,624
88	1817	1,849
89	1818	1,838
90	1819	1,805
91	1820	1,863
92	1821	1,705
93	1822	1,731
94	1823	1,770
95	1824	1,709

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cohort #	years	# obs
96	1825	1,845
97	1826	1,685
98	1827	1,760
99	1828	1,746
100	1829	1,795
101	1830	1,890
102	1831	1,692
103	1832	1,693
104	1833	1,786
105	1834	1,775
106	1835	1,785
107	1836	1,835
108	1837	1,873
109	1838	1,860
110	1839	1,931
111	1840	2,069
112	1841	1,958
113	1842	2,001
114	1843	1,972
115	1844	1,951
116	1845	2,048
117	1846	1,953
118	1847	1,966
119	1848	2,067
120	1849	1,882
121	1850	2,097
122	1851	2,022
123	1852	2,079
124	1853	1,901
125	1854	1,956
126	1855	2,056
127	1856	2,134

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cohort #	years	# obs
128	1857	2,055
129	1858	2,292
130	1859	2,239
131	1860	2,286
132	1861	2,212
133	1862	2,304
134	1863	2,300
135	1864	2,267
136	1865	2,296
137	1866	2,273
138	1867	2,245
139	1868	2,410
140	1869	2,367
141	1870	2,301
142	1871	2,128
143	1872	2,288
144	1873	2,296
145	1874	2,286
146	1875	2,381
147	1876	2,327
148	1877	2,267
149	1878	2,309
150	1879	2,349
Total		297,651

G Estimation of the Gompertz-Makeham Law

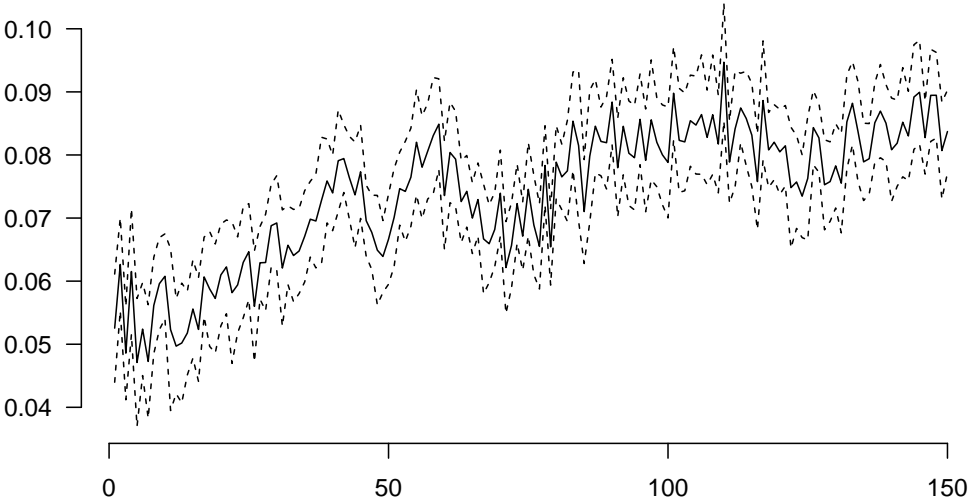


Figure A.7: Estimated $\hat{\alpha}$

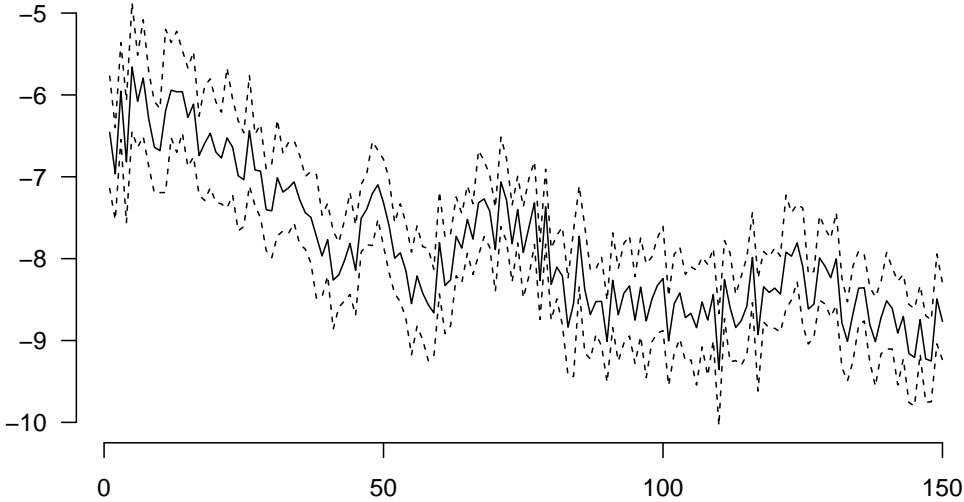


Figure A.8: Estimated $\hat{\rho}$

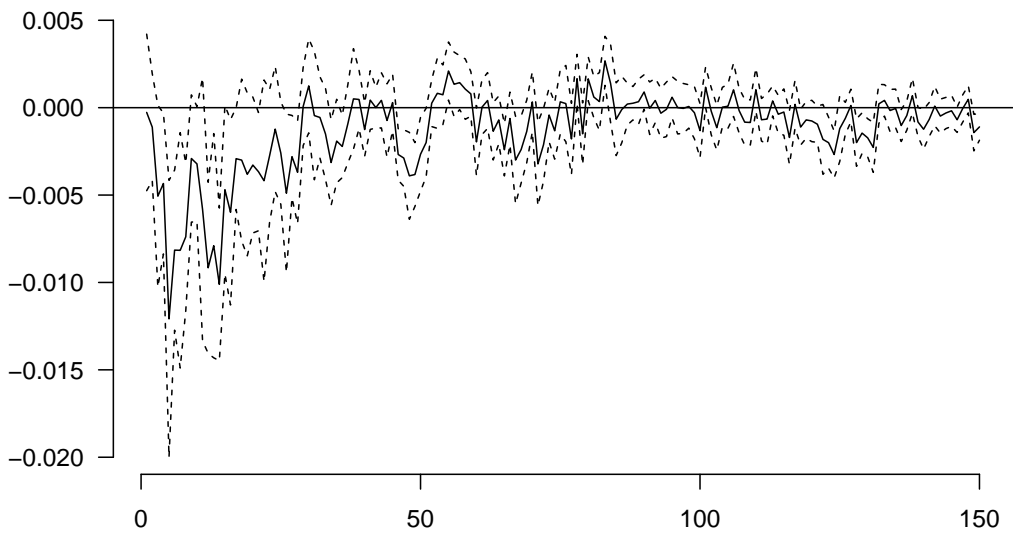


Figure A.9: Estimated \hat{A}

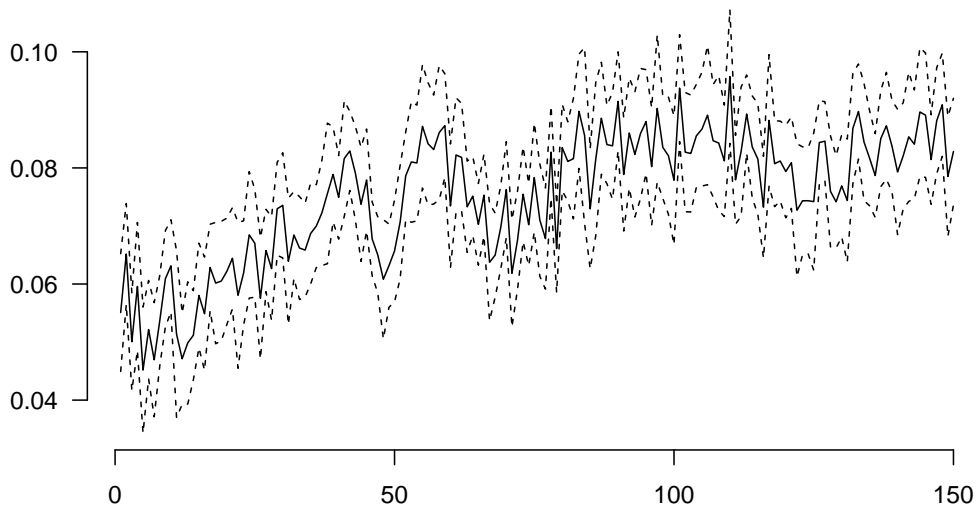


Figure A.10: Estimated \hat{a} - Notoriety Bias Corrected

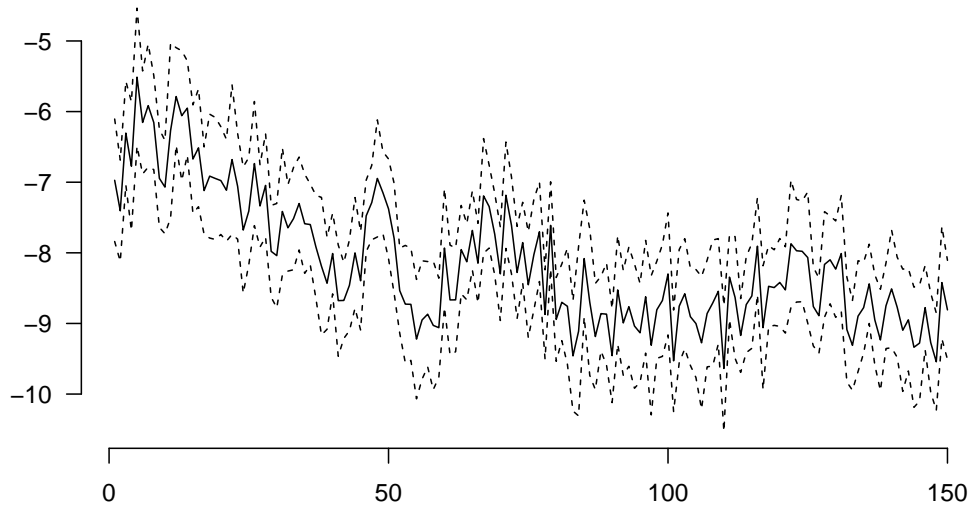


Figure A.11: Estimated $\hat{\rho}$ - Notoriety Bias Corrected

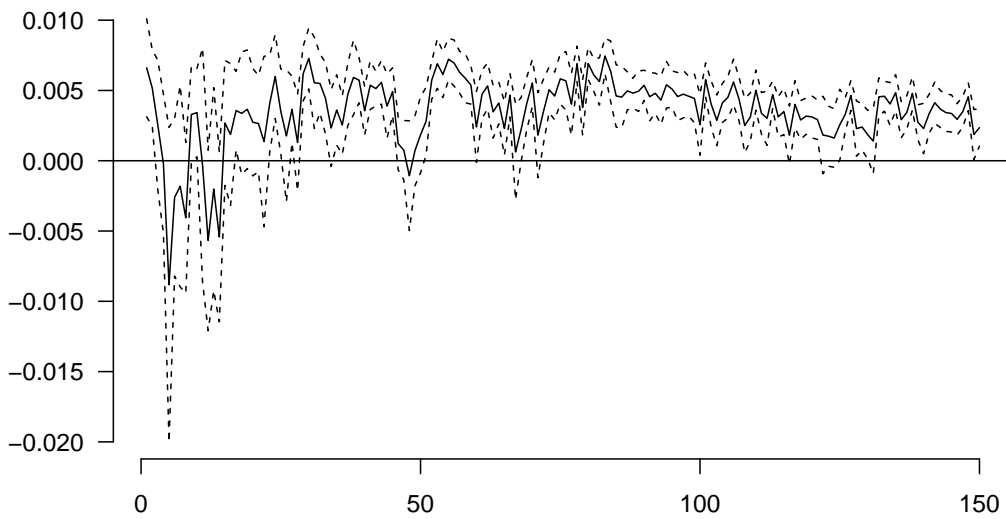


Figure A.12: Estimated \hat{A} - Notoriety Bias Corrected

References

- A'Hearn, Brian, Joerg Baten, and Dorothee Crayen. 2009. "Quantifying quantitative literacy: age heaping and the history of human capital." *Journal of Economic History* 69:783–808.
- De Moor, Tine, and Jaco Zuijderduijn. 2013. "The Art of Counting: Reconstructing Numeracy of the Middle and Upper Classes on the Basis of Portraits in the Early Modern Low Countries." *Historical Methods: A Journal of Quantitative and Interdisciplinary History* 46 (1): 41–56.