Historical Analysis of National Wellbeing Using Millions of Digitized Books

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Subjective Wellbeing and Gross Domestic Product

- Subjective wellbeing (or "happiness") has played a minor role in the development and application of economic policy in the past
- Growing literature on international patterns of subjective wellbeing.
- Several nations including the UK, Australia, China, France and Canada now collect subjective wellbeing data to use alongside GDP in national measurement exercises.

A Parallel with GDP

- Development of GDP in the 1930s immediately following the Great Depression; Simon Kuznets (early developer) had different ideas about GDP (e.g., shouldn't include military spending or dis-services).
- Problems with GDP as a way to capture wellbeing:
    - Enviromnent: BP Deep Horizons oil spill increased US GDP.
    - Leisure is not included: wealthier people may choose to "buy" leisure but then income "falls".
    - Other issues: exchange rates, goods/output change over time (centuries), informal and Black economies.

A Parallel with GDP

- But these issues do not forestall the need to roll back GDP figures to better understand the evolution of national income and its drivers.

- Maddison Historical GDP Project rolls back GDP to the early 19th century, Broadberry et al going back much further for Britain and the Netherlands.

Our Approach

- Our primary objective is to produce a workable proxy for subjective wellbeing going back to 1800, which would enable direct comparisons with GDP over that period.
- Our methods rely on the digitization of books, available in the Google Books corpus.
- We elected to start in 1800 because the number of digitalized books is too small before.
- 1776 is the date of the American Declaration of Independence, one of the most famous of all historical documents to specifically reference happiness.

Outline

Word Norms or *Valence*

- The approach we take here is a common approach among the studies Inferring public mood and relies on affective word norms to derive sentiment from text
- In a study of 17 million blog posts, (Nguyen et al, 2010) found that a simple calculation based on the weighted affective ratings of words was highly effective (70% accuracy) at predicting the mood of blogs compared against the groundtruth provided by the bloggers

Language Corpus Data

- The language corpora we used is the *Google Books Ngram* Corpus https://books.google.com/ngrams
- Overall, this data represents about 6% of all books ever published.
- The corpus is based on a digitalised database of physically published books, which was developed as part of the Google Books programme.
- We analysed data for 6 languages, English (British), English (American), German, Italian, Spanish, French.
- There are no word norms available for Chinese, Hebrew and Russian

Word Norms

- In order to assess the valence of individual words, we used the largest available sets of existing word valence rating norms for each language.

- Word valence rating norms generally ask participants to rate each word from a list on how positive or negative they perceive a word to be.

- To allow for comparison across languages, all of our valence norms use a subset of words. There is a list of a thousand words that served as the basis for developing valence ratings for multiple languages through several independent studies.

Affective Norms for Different Languages

- For English, ANEW contains about 10,000 words, all rated on a 1 to 9 valence scale.

- For German, we used the Affective norms for German sentiment terms. This is a list of 1003 words, a German translations of the ANEW list. The valence ratings were collected on a -3 to +3 scale. The mean values were adjusted to reflect a 1 to 9 scale.

- the French and Spanish norms were also adaptations of the ANEW. These contained 1031 and 1034 words respectively. Both used a 1 to 9 points scale.

- For Italian, we used an adaptation of the ANEW norms containing 1121 Italian words, based on the ANEW material on a 1 to 9 scale.

Valence and Words in different languages

ENGLISH	VALENCE	GERMAN	VALENCE	FRENCH	VALENCE	ITALIAN	VALENCE	SPAIN	VALENCE
aardvark		Misshandlun	-2.6	abeille	4.22	abbaglio	3.94	abandonado	1.68
abalone	6.26	Unfall	-2.35	abonné	4.53	abbandonato	2	abejas	3.18
abandon	5.3	Erfolg	2.1	abricot	6.55	abbondanza	6.82	aborto	2.8
abandonment	2.84	aktivieren	0.81	absent	3.42	abbraccio	7.7	abrasador	2.46
abbey	2.63	bewundert	1.44	abstrait	4.72	abete	6.17	abrazo	8.13
abdomen	5.85	anbetungswi	0.4	accordéon	5.7	abitante	5.67	abrumado	2.9
abdominal	5.43	Zuneigung	1.9	acide	3.47	abitazione	6.46	absurdo	3.8
abduct	4.48	ängstlich	-1.38	agneau	6.35	abito	7.27	abundancia	6.8
abduction	2.42	aggressiv	-1.6	agréable	8.29	abitudini	4.91	aburrido	2.33
abide	2.05	Qual	-2.7	aïde	7.08	aborto	2.06	accidente	1.32
abiding	5.52	Alarm	-2.55	aigle	6.53	abuso	1.74	ácido	3.41
ability	5.57	allein	-1.8	aiguille	3.9	accettazione	5.79	acogedor	7.64
abject	7	entfernt	-0.6	ail	4.22	accogliente	8.03	acontecimier	5.99
ablaze	4	Ehrgeiz	1.5	aile	6.05	accomodante	6.4	acre	4.23
able	5.15	Krankenwage	-0.79	aisance	7.26	accordo	6.71	activar	6
abnormal	6.64	Engel	2.15	album	6.34	acqua	7.78	acuerdo	7.24
abnormality	3.53	Ärger	-1.9	alcool	5.64	adorabile	7.33	acurrucarse	6.98
abode	3.05	verärgert	-1.27	algèbre	3.87	adulto	5.78	adicto	2.41
abolish	5.28	verängstigt	-1.56	allégorie	5.42	aereo	6.56	adinerado	6.21
abominable	3.84	verängstigt	-1.56	alligator	4.05	affamato	4.74	admirado	7.33
abomination	4.05	ärgern	-1.22	allumette	5.32	affascinare	7.97	adorable	7.48
abort	2.5	gespannt	0.73	ambition	7.6	affaticato	3.73	adulto	5.68
abortion	3.1	Beweglichkei	1.53	ambulance	3.22	affetto	7.48	afectar	3.48
abracadabra	2.58	Armee	-2.1	âme	7.12	afflizione	1.94	afecto	8.1
abrasive	5.11	erregt	1	amer	2.8	affogare	1.79	afianzar	5.93
abreast	4.26	arrogant	-1.93	ami	7.94	aggressione	2.53	afligido	1.96
abrupt	4.62	ermorden	-2.21	amitié	8.38	aggressivo	3.48	afortunado	7.71

Language Average Valence Computation

- For each language we compute the weighted valence score, $Valence_t$, for each year, t, using the valence, v for each word, j, as follows,
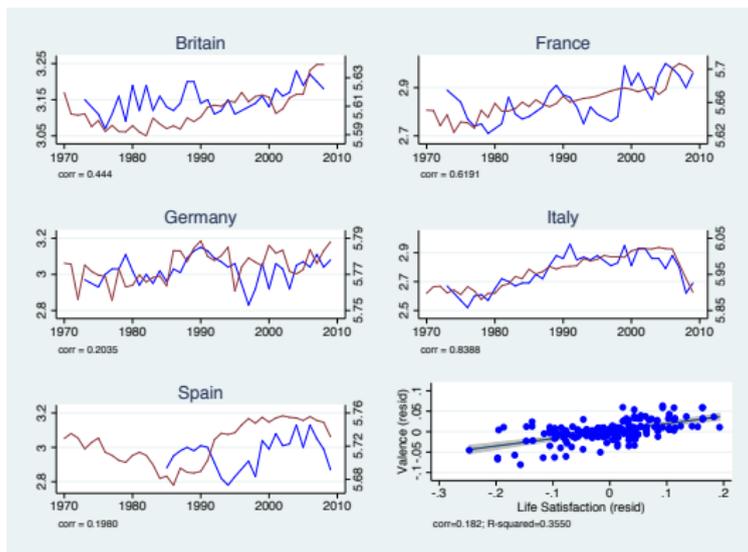
$$Val_{i,t} = \sum_{j=1}^{n} v_{j,i} p_{j,i,t};$$

- Note that $v_{j,i}$ is the valence for word j as found in the appropriate valence norms for language i, and $p_{j,i,t}$ is the proportion of word j in year t for the language i.

How to Interpret the Index

- Think about the book market as highly competitive (lots of potential writers and publishers): publishers "match" books to demand.
- It could be that publishers match happy people to happy books and the opposite?
- It could be that writers are inspired by periods and happy period inspires happy books and the opposite?
- We will try to answer this question by comparing the available data on SWB with word-valence based index

Valence and Existing data of Life Satisfaction
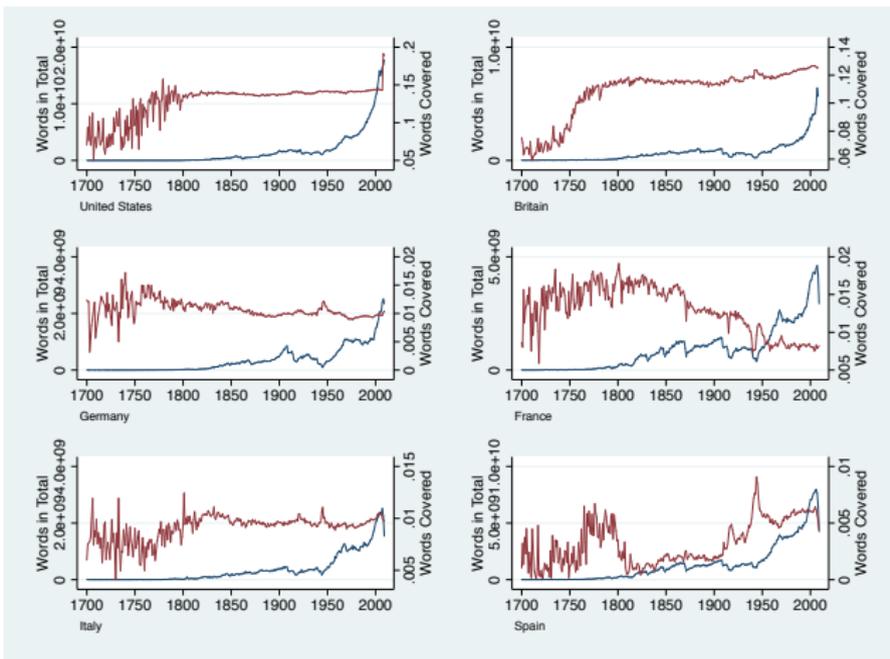
Valence Predicts Aggregate Life Satisfaction

Table: Average life satisfaction per country and year is the dependent variable.

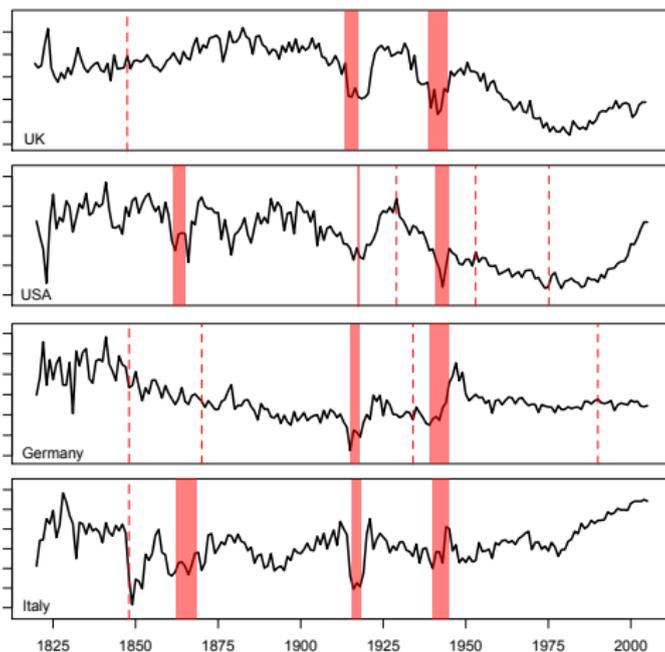|	1	2	3	4	5	6
	Baseline	Until 2000	w/o Sp.and Fr.	Trends	+GDP	Year FE
	b/se	b/se	b/se	b/se	b/se	b/se
Valence	1.9554***	1.6941***	2.1696***	1.5549***	0.7180**	1.6107***
	(0.2221)	(0.3093)	(0.2339)	(0.3408)	(0.3499)	(0.2784)
Log GDP					0.8243***	0.1452
					(0.1537)	(0.1300)
Words Covered	0.9816	−0.0037	−0.4491	8.7147	−0.1693	−15.6331*
	(6.2645)	(9.1248)	(5.7111)	(15.0425)	(13.9245)	(8.5557)
Country FE	Yes	Yes	Yes	Yes	Yes	Yes
Country Specific Trend	No	No	No	Yes	Yes	No
Year FE	No	No	No	No	No	Yes
r2	0.358	0.227	0.501	0.387	0.485	0.645
N	163	119	104	163	163	163

Data Concerns

- Long-run biases might emerge from country-specific factors such as culture, language, religion and demographics (immigration, population age structure). We can control these to some extent through country fixed effects.

- Literacy was lower in the past, Language different. We control for education, trends, year fixed effect.

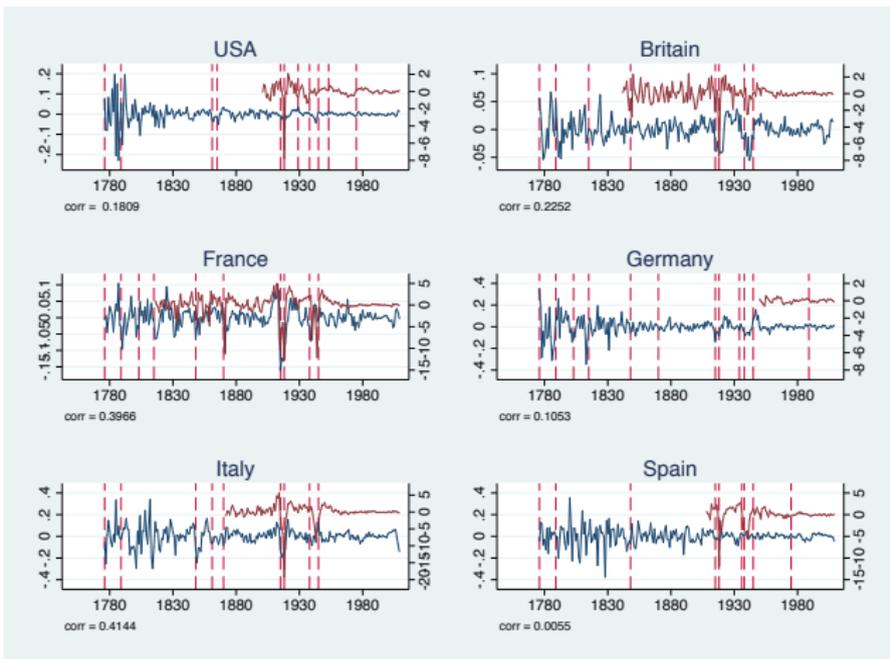- Freedom of the press, we control for democracy.

Words and Words Covered.

A Time-Series Plot of the Valence Index Over the Period 1820-2009

HP-filtered Life Expectancy and Valence Index

Historical Determinants of the Valence Index from 1820 to 2009.

Table: The countries included are Germany, Italy, the UK and the United States

	1 Year FE b/se	2 CS Trends b/se
(log) GDP(t-3)	0.0821**	0.0517*
	(0.0174)	(0.0213)
Life Expectancy(t)	0.0036**	0.0016
	(0.0008)	(0.0014)
Internal Conflict(t)		−0.0190**
		(0.0049)
World Covered(t)	Yes	Yes
Democracy(t)	Yes	Yes
Education Inequality(t)	Yes	Yes
Year FE	Yes	No
Country-Specific Trends	No	Yes
r2	0.736	0.494
N	412	412

Summary

- Average Word Valence of a language predict country aggregate Subjective Wellbeing of the corresponding country
- Valence Index positively correlates with Life Expectancy, GDP and other relevant variables

Thank you !

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