# Measuring the Happiness, Health, and Stories of Populations

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Complex Networks | @networksvox CSYS/MATH 303, Spring, 2019

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e Computational Story Lab

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# Outline

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### Key papers:

"Measuring the happiness of large-scale written expression: Songs, blogs, and presidents." C, Dodds and Danforth, Journal of Happiness Studies, **11**, 441–456, 2009. <sup>[9]</sup>

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"Temporal patterns of happiness and information in a global social network: Hedonometrics and Twitter" Dodds et al., PLoS ONE, **6**, e26752, 2011.<sup>[11]</sup>

"Positivity of the English language" C, Kloumann et al., PLoS ONE, **7**, e29484, 2012.<sup>[23]</sup>



"Sentiment analysis methods for understanding large-scale texts: A case for using continuum-scored words and word shift graphs" <sup>(2)</sup>, Reagan et al., EPJ Data Science, **6**, , 2017.<sup>[31]</sup> COcoNuTS @networksvox Happiness

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# **Economics**, Schmeconomics

### Alan Greenspan (September 18, 2007):

"I've been dealing with these big mathematical models of forecasting the economy ...

If I could figure out a way to determine whether or not people are more fearful or changing to more euphoric,

I don't need any of this other stuff.

I could forecast the economy better than any way I know."



http://wikipedia.org

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# **Economics**, Schmeconomics

### Greenspan continues:

"The trouble is that we can't figure that out. I've been in the forecasting business for 50 years. I'm no better than I ever was, and nobody else is. Forecasting 50 years ago was as good or as bad as it is today. And the reason is that human nature hasn't changed. We can't improve ourselves."

### Ion Stewart:

"You just bummed the @\*!# out of me."



wildbluffmedia.com



From the Daily Show C (September 18, 2007; @5:13) The full inteview is here  $\mathbf{C}$ .

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# This is a Collateralized Debt Obligation:

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# Robert Kennedy on the Gross Domestic Product (GDP) in 1968:

"It measures everything except that which makes life worthwhile. And it can tell us everything about America except why we are proud that we are Americans." <sup>[21, 20]</sup>

# Basic Science $\simeq$ Describe + Explain:

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But also:

🚳 "X-rays will prove to be a hoax."

discovered in physics now. All that remains is more and more

Lord Kelvin (possibly):

🚓 "To measure is to know."

cannot improve it."

🚳 "lf you cannot measure it, you

🚳 "There is nothing new to be

precise measurement."

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### A brief history of measuring time:

- 🚳 Megaliths for Big Time
- 🗞 Sundials, 1500 BC, Egypt (solid for over 2000 years)
- Escapements (200s), Hourglasses (1300s?), Pendulum clocks (Galileo, 1500s)

### Chronometers, 1700s:



"Longitude: The True Story of a Lone Genius Who Solved the Greatest Scientific Problem of His Time, publisher=Bloomsbury Publishing, US" by Dava Sobel (2007). [33]

Billionths of a second accuracy: Atomic clocks (Lord Kelvin, 1879)

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### Measuring temperature was thought impossible:

The properties measured by our instruments usually begin as subjective judgments. Temperature is a good example. People were aware of variations in temperature long before there were any objective measurements of temperature. Judgments of temperature are imperfectly correlated among different persons, or even the same person at different times, depending on the humidity, the person's activity level and age, surrounding air currents, and so on. The idea that anything as subtle and complex as all the manifestations of changes in temperature could be measured and quantified on a single numerical scale was scoffed at as impossible, even by the leading philosophers of the sixteenth century.

The first thermometer invented by Galileo in 1592 did not go far in dispelling the notion that temperature was inherently unmeasurable, because the earliest thermometers, for about their first hundred years, were so imperfect as to make it possible for those who wished to do so to argue that no one could ever succeed in measuring temperature. Temperature was then confounded with all the subtleties of subjective judgment, which easily seem incompatible with a single numerical scale of measurement. How could the height of a column of mercury in a glass tube possibly reflect the rich varieties of temperature—damp cold, dank cold, frosty cold, crisp cold, humid heat, searing heat, scalding heat, dry heat, feverish heat, prickly heat, and so on?

From "Bias in Mental Testing", Arthur Jensen, 1980<sup>[17]</sup> per @Silver/Vulpes 7: Also: Inventing Temperature, Hasok Chang, 2004<sup>[3]</sup> COcoNuTS @networksvox Happiness

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### Measuring temperature was thought impossible:

The early thermometers were inconsistent, both with themselves and with each other. Because they consisted of open-ended glass tubes, they were sensitive to changes in barometric pressure as well as to temperature. And there were problems of calibration, such as where to locate the zero point and how to divide the column of mercury into units. It was believed, incorrectly, that all caves had the same temperature, so thermometers were calibrated in caves. The freezing and boiling points of water were also used in calibration, but, as these vary with impurities in the water and the barometric pressure, the calibration of different thermometers at different times and places resulted in thermometers that failed to correlate perfectly with one another in any given instance. They lacked reliability, as we now would say.

All the while, no one knew what temperature is in a theoretical or scientific sense. There was no theory of thermodynamics that could explain temperature phenomena and provide a complete scientific rationale for the construction and calibration of thermometers. Yet quite adequate and accurate thermometers, hardly differing from those we use today, were eventually developed by the middle of the eighteenth century. Thus the objective measurement of temperature considerably preceded the development of an adequate theory of temperature and heat, and necessarily so, as the science of thermodynamics could not possibly have developed without first having been able to quantify or measure the temperatures of liquids, gasses, and other substances independently of

From "Bias in Mental Testing", Arthur Jensen, 1980<sup>[17]</sup> per @Silver/Vulpes 7: Also: Inventing Temperature, Hasok Chang, 2004<sup>[3]</sup> COcoNuTS @networksvox Happiness

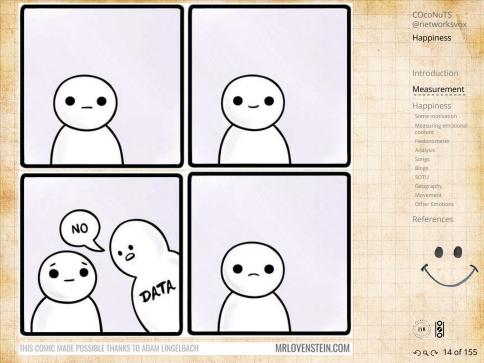
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# What matters and what's measurable:

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### Panometer—Three kinds of lexical meters:



- Principled lexical meters:
   The Hedonometer.
  - The Lexicocalorimeter.
- 2. Ground truth lexical meters:
  - lnsomniometer.
  - Hangoverometer.
- 3. Bootstrap lexical meters:
  - Boredometer.
    - Hashtagometers.

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# Measuring Happiness:

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Socrates et al.: eudaimonia <sup>[18]</sup>

### Bentham: hedonistic calculus

Jefferson: ...the pursuit of happiness



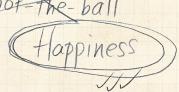
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### Early drafts:

that among these are: Life,

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and



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# Happiness:

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Even the odd modern economist is happy:

"Happiness" by Richard Layard <sup>[24]</sup>

MAPPINESS

BICHARD LAYARD

[amazon]

What makes us happy?—Layard's summary:

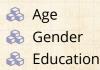
Dominant factors:





Community and Friends

### Unimportant factors:



Inherent
 intelligence
 Looks

Health

Personal

Freedom

Personal Values

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### Desiring happiness—not just for boffins:

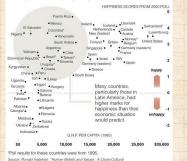
- Average people routinely report being happy is what they want most in life <sup>[24, 25, 8]</sup>
- And it matters: "Happy people live longer:..." Survey by Diener and Chan. [8]

#### A Plateau of Happiness

Sourcebook Based on the 1999-2002 Values Surveys

#### A country's wealth may not always dictate the happiness of its people.

As part of the World Values Survey project, inhabitants of different countries and territories were asked how happy or satisfied they were. Below is a sampling of happiness rankings, along with economic status.



National indices of well-being:



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# Some easy knocks:

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State of Mind And Other Aphorisms

The Passionate

"The Passionate State of Mind: And Other Aphorisms" **a** C by Eric Hoffer (1954).<sup>[16]</sup>

"The search for happiness is one of the chief sources of unhappiness."

# Some easy knocks:

### Colbert: "Happiness is totally overrated ..."

### "Happiness is for the weak."



### Full interview with Jennifer Senior here 🕝 (2014/02/03)

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## 30 Rock, S7E8:

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### 30 Rock, S7E8:



JD: "Before she died, Colleen told me—she just wants me to be happy. 'I just want you to be happy.' You know who you say that to? A loser. Someone who can't hope for anything more in life than just being happy. You say that to someone who has disappointed you."

LL: "Jack."

JD: "No. It's perfect. She's a genius. One last twist of the knife. Well, thank you for coming, Lemon, but I better get going. The funeral is tomorrow. Colleen wanted to be buried before the rest of the family found out and sold her body to a haunted house. And, of course, I get to eulogize Colleen at the service. One more chance to disappoint her as she looks up at me from her throne In hell.

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### Meaning rather than happiness:



"Mindfulness in Plain English" **3** C by Henepola Gunaratana (1992).<sup>[15]</sup>



"Flow" **3** by Mihaly Csikszentmihalyi (1990). <sup>[6]</sup>

Can we measure Flow in a big data way?
 Maybe drops in social media usage indicate people are doing okay?

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# **Emotional content**

### So how does one measure

- 1. happiness?
- 2. levels of other emotional states?

### Just ask people how happy they are.

Experience sampling<sup>[5, 7, 6]</sup> (Csikszentmihalyi et al.)
 Day reconstruction<sup>[19]</sup> (Kahneman et al.)

### But self-reporting has some drawbacks:

relies on memory and self-perception
 induces misreporting <sup>[26]</sup>
 costly



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# Happiness, attention, and doing:

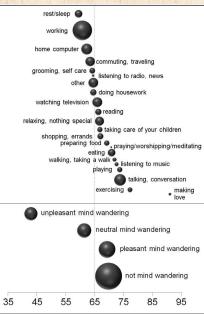


Fig. 1. Mean happiness reported during each activity (top) and while mind wandering to unpleasant topics, neutral topics, pleasant topics or not mind wandering (bottom). Dashed line indicates mean of happiness across all samples. Bubble area indicates the frequency of occurrence. The largest bubble ("not mind wandering") corresponds to 53.1% of the samples, and the smallest bubble ("praying/worshipping/meditating") corresponds to 0.1% of the samples.

Killingsworth and Gilbert, Science, 2010<sup>[22]</sup> Some mouvailon Measuring emotional content Hedonometer Analysis Songs Blogs SOTU Geography Movement

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### We'd like to build an 'hedonometer':



An instrument to 'remotely-sense' emotional states and levels, in real time or post hoc.

### Ideally:



- \lambda Transparent
  - Fast
- Based on written expression
- 🚳 Uses human evaluation

Non-reactive Complementary to self-reported measures Improvable 23

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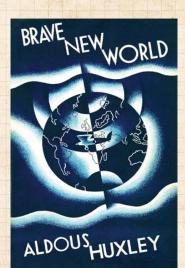




# We don't want to end up here:

George Orwell 10 our a novel

Science



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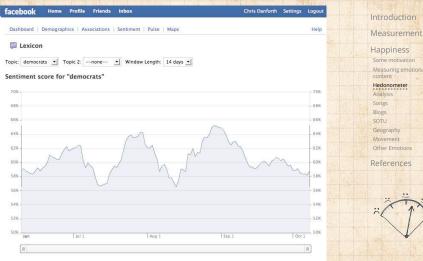
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# Or here: Facebook Lexicon Sentiment Analysis (2008)



Sentiment shows the percentage of posts that are positive vs. negative about the topic. For example, the phrase "I love Facebook" would be identified as a positive sentiment. An 80% score means that 80% of the sentiment is positive and 20% is negative.

Drag the ends of the slider to zoom into a specific period of time.

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# Facebook Lexicon Sentiment Analysis

### Binary decision on emotional content



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Limitation: Sentiments are classified as either positive or negative.

4 'I like Sarah Palin' given same score as 'Sarah Palin's voice fills me with unbridled joy!'



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### Measuring Emotional Content

- Idea: Build on measures of the emotional content of individual words.
- Osgood et al. (1957)<sup>[30]</sup> identified a basis of three psychological variables as semantic differentials:
  - ✔ Valence: bad ↔ good
     ✔ Arousal: passive ↔ active
  - ♥ Dominance: weak ↔ strong
  - Also often: Evaluation, Activity, and Potency.

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# **ANEW** study

ANEW = "Affective Norms for English Words"

- Study: participants shown lists of isolated words
   Asked to grade each word's valence, arousal, and dominance level
- Integer scale of 1–9
- N = 1034 words—previously identified as bearing emotional weight
- Participants = College students (\*cough\*)
- Results published by Bradley and Lang (1999)<sup>[2]</sup>

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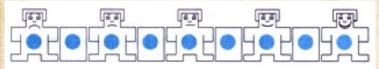
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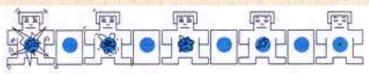
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# 1999 ANEW study—three 1–9 scales: <sup>[2]</sup>

valence:



arousal:

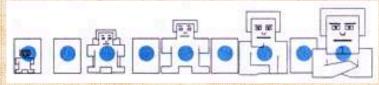


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### dominance:



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# ANEW study:

### Valence = Happiness:

Valence scale presented to participants as a 'happy-unhappy scale.'

Participants were further told:

"At one extreme of this scale, you are happy, pleased, satisfied, contented, hopeful. ...

The other end of the scale is when you feel completely unhappy, annoyed, unsatisfied, melancholic, despaired, or bored." COcoNuTS @networksvox Happiness

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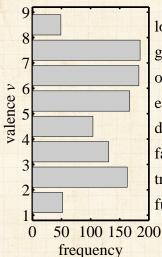
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#### ANEW study words—examples



love/paradise/triumphant glory/luxury/trophy optimism/pancakes/church engine/paper/street derelict/neurotic/vanity fault/corrupt/lawsuit trauma/hostage/disgusted funeral/rape/suicide

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#### Analysing text:

Simplest measure for a text:

$$\theta_{\mathsf{avg}} = \sum_{i=1}^N p_i \theta_i$$

where  $p_i$  is fractional abundance of word i and  $\theta$  is average valence, arousal, or dominance for word i. So Focus on happiness (valence),  $\theta = h$ . Average happiness typically falls between 5 and 7. COcoNuTS @networksvox Happiness

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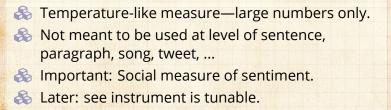
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# Measuring the perceived happiness of a text:







#### Daft Punk's "Around the World"

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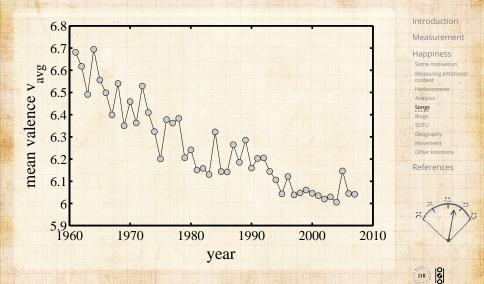
Around the world, around the world Around the world, around the world Around the world, around the world Around the world, around the world

Around the world, around the world Around the world, around the world Around the world, around the world Around the world, around the world

Magic: Low entropy, high energy.

# Song Lyrics—average happiness

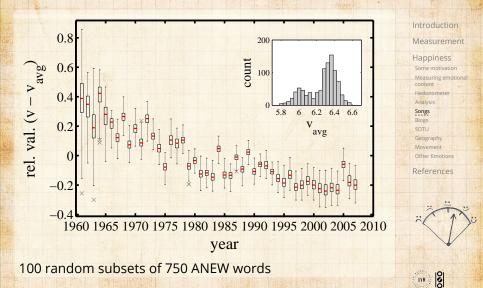
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### Song Lyrics—measurement robustness

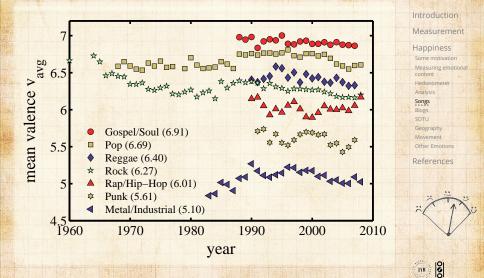
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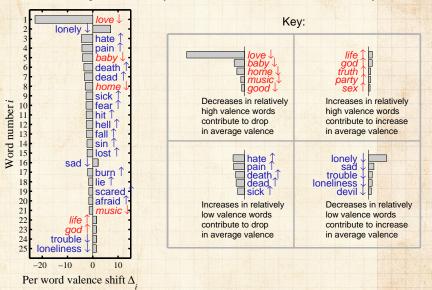
## Song Lyrics—average happiness of genres:

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#### Happiness Word Shift Graph (early version):

Per word drop in valence of lyrics from 1980–2007 relative to valence of lyrics from 1960



# Word data shift details:

#### Given two texts $T_{ref}$ and $T_{comp}$ :

- Evident question: Which words contribute the most to this change?
- Break difference down by contributions from individual words:

$$\delta h_{\mathrm{avg},\mathrm{i}} = \frac{100}{\left|h_{\mathrm{avg}}^{(\mathrm{comp})} - h_{\mathrm{avg}}^{(\mathrm{ref})}\right|} \underbrace{\left[h_{\mathrm{avg}}(w_i) - h_{\mathrm{avg}}^{(\mathrm{ref})}\right]}_{+/-} \underbrace{\left[p_i^{(\mathrm{comp})} - p_i^{(\mathrm{ref})}\right]}_{\uparrow/\downarrow}$$

Must have:  $\sum_i \delta h_{\text{avg},i} = \pm 100$  Rank words by  $|\delta h_{\text{avg},i}|$ 

# Word data shift details:

$$h_{\mathrm{avg}}^{(\mathrm{comp})} - h_{\mathrm{avg}}^{(\mathrm{ref})} = \sum_{i=1}^{N} h_{\mathrm{avg}}(w_i) p_i^{(\mathrm{comp})} - \sum_{i=1}^{N} h_{\mathrm{avg}}(w_i) p_i^{(\mathrm{ref})} + \sum_{i=1}^{N} h_$$

$$= \sum_{i=1}^{N} h_{\text{avg}}(w_i) \left[ p_i^{(\text{comp})} - p_i^{(\text{ref})} \right]$$

$$= \sum_{i=1}^{N} \left[ h_{\text{avg}}(w_i) - h_{\text{avg}}^{(\text{ref})} \right] \left[ p_i^{(\text{comp})} - p_i^{(\text{ref})} \right]$$

#### where

$$\sum_{i=1}^{N} h_{\mathrm{avg}}^{(\mathrm{ref})} \left[ p_i^{(\mathrm{comp})} - p_i^{(\mathrm{ref})} \right] = h_{\mathrm{avg}}^{(\mathrm{ref})} \sum_{i=1}^{N} \left[ p_i^{(\mathrm{comp})} - p_i^{(\mathrm{ref})} \right]$$

$$= h_{\text{avg}}^{(\text{ref})}(1-1) = 0.$$

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- + $\uparrow$ : Increased usage of relatively positive words—If a word is happier than text  $T_{ref}$  (+) and appears relatively more often in text  $T_{comp}$  ( $\uparrow$ ), then the contribution to the difference  $h_{avg}^{(comp)} - h_{avg}^{(ref)}$  is positive;
- $-\downarrow$ : Decreased usage of relatively negative words—If a word is less happy than text  $T_{ref}$  (–) and appears relatively less often in text  $T_{comp}$  ( $\downarrow$ ), then the contribution to the difference  $h_{avg}^{(comp)} h_{avg}^{(ref)}$  is also positive;
- +↓: Decreased usage of relatively positive words—If a word is happier than text  $T_{ref}$  (+) and appears relatively less often in text  $T_{comp}$  (↓), then the contribution to the difference  $h_{avg}^{(comp)} h_{avg}^{(ref)}$  is negative; and
- $-\uparrow$ : Increased usage of relatively negative words—If a word is less happy than text  $T_{ref}$  (–) and appears relatively more often in text  $T_{comp}$  ( $\uparrow$ ), then the contribution to the difference  $h_{avg}^{(comp)} - h_{avg}^{(ref)}$  is also negative.

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UVN S

Top 50 of  $\simeq$  20,000 artists:

Rank Artist Rank Artist  $h_{\text{avg}}$  $h_{avg}$ All-4-One 7.15 26 Sarah Connor 6.86 2 Luther Vandross 27 Darlene Zschech 6.86 7.12 3 S Club 7 7.05 28 Mary | Blige 6.86 K Ci & lolo 7.04 Steve Miller Band 4 29 6.86 5 7.04 New Edition 6.86 Perry Como 30 6 Diana Ross & The Supremes 7.03 31 Mandy Moore 6.86 7 **Buddy Holly** 32 7.02 Alicia Keys 6.85 Faith Evans 33 Cher 8 7.01 6.85 9 The Beach Boys 7.01 34 Modern Talking 6.85 35 10 Jon B 6.98 Mario 6.84 Dru Hill 6.96 36 Aretha Franklin 11 6.84 12 **Farth Wind & Fire** 6.95 37 Jessica Simpson 6.84 13 Ashanti 6.95 38 112 6.84 14 Otis Redding 6.93 39 6.83 Backstreet Boys Faith Hill **Billy Gilman** 15 6.93 40 6.83 16 NSync 6.93 41 B2K 6.82 17 42 The Supremes 6.91 Stevie Wonder 6.82 The Partridge Family 18 6.91 43 John Legend 6.81 **Ricky Nelson** 19 Kelly Price 6.89 44 6.79 20 Tamia 6.89 45 Lionel Richie 6.79 21 46 6.79 Avant 6.88 98 Degrees 22 lennifer Lopez 6.88 47 Bovzone 6.79 23 Vanessa Williams 6.87 48 Gerald Levert 6.79 24 6.78 Babyface 6.87 49 Nat King Cole 25 6.78 E Rotic 6.87 50 Margues Houston

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(in the second s

(criteria:  $\geq$  50 songs and  $\geq$  1000 ANEW words)

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Bottom 50 of  $\simeq$  20,000 artists:

Rank

1 2

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Twiztid

Evanescence

|                          |                    |      |                    |                    | a strate           |
|--------------------------|--------------------|------|--------------------|--------------------|--------------------|
| Artist                   | $h_{\mathrm{avg}}$ | Rank | Artist             | $h_{\mathrm{avg}}$ | Happir             |
| Slayer                   | 4.80               | 26   | Nine Inch Nails    | 5.34               |                    |
| Misfits                  | 4.88               | 27   | Sevendust          | 5.34               |                    |
| Staind                   | 4.93               | 28   | Annihilator        | 5.35               | Introdu            |
| Slipknot                 | 4.98               | 29   | Biohazard          | 5.36               | Measu              |
| Darkthrone               | 4.98               | 30   | Insane Clown Posse | 5.36               | Happir             |
| Death                    | 5.02               | 31   | Megadeth           | 5.36               | Some mo            |
| Black Label Society      | 5.05               | 32   | Manowar            | 5.37               | Measurin           |
| Pig                      | 5.08               | 33   | Zebrahead          | 5.38               | content<br>Hedonon |
| Voivod                   | 5.14               | 34   | Danzig             | 5.39               | Analysis           |
| Fear Factory             | 5.15               | 35   | Acid Drinkers      | 5.40               | Songs              |
| lced Earth               | 5.16               | 36   | Dag Nasty          | 5.40               | Blogs              |
| Simple Plan              | 5.16               | 37   | Iron Maiden        | 5.40               | Geograph           |
| Machine Head             | 5.17               | 38   | Flotsam And Jetsam | 5.41               | Moveme             |
| Metallica                | 5.19               | 39   | Powerman 5000      | 5.42               | Other Em           |
| Dimmu Borgir             | 5.20               | 40   | Anthrax            | 5.43               | Refere             |
| Mudvayne                 | 5.21               | 41   | Rhapsody           | 5.43               |                    |
| Linkin Park              | 5.22               | 42   | Korn               | 5.43               |                    |
| Papa Roach               | 5.22               | 43   | Rage               | 5.44               |                    |
| Audioslave               | 5.24               | 44   | Accept             | 5.45               | :1                 |
| Rage Against The Machine | 5.24               | 45   | Esham              | 5.46               | 1                  |
| Cradle Of Filth          | 5.25               | 46   | Blind Guardian     | 5.46               |                    |
| Dark Tranquility         | 5.26               | 47   | White Zombie       | 5.47               |                    |
| Jack Off Jill            | 5.28               | 48   | Helloween          | 5.50               |                    |
|                          |                    |      |                    |                    |                    |

WASP

Green Day

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iness

urement

5.50

5.50

5.33 (criteria:  $\geq$  50 songs and  $\geq$  1000 ANEW words)

5.30

49

50

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| Text:                            | $h_{avg}$ | Words with a similar score:                       |
|----------------------------------|-----------|---|
| Soul/Gospel                      | 6.9       | chocolate (6.88), leisurely (6.88), penthouse     |
| lyrics <sup>[10]</sup>           |           | (6.81)  |
| Pop lyrics <sup>[10]</sup>       | 6.7       | dream (6.73), honey (6.73), sugar (6.74)          |
| Dante's Paradise <sup>[</sup> ?] | 6.5       | muffin (6.57), rabbit (6.57), smooth (6.58)       |
| Tweets, 9/9/2008 to              | 6.4       | thought (6.39), face (6.39), blond (6.42)         |
| 12/31/2010                       |           |   |
| Rock lyrics <sup>[10]</sup>      | 6.3       | church (6.28), tree (6.32), air (6.34)            |
| Enron Emails <sup>[?]</sup>      | 6.2       | clouds (6.18), alert (6.20), computer (6.24)      |
| State of the Union               | 6.1       | grass (6.12), idol (6.12), bottle (6.15)          |
| Messages <sup>[10]</sup>         |           |   |
| New York Times                   | 6.0       | hotel (6.00), tennis (6.02), wonder (6.03)        |
| (1987–2007) <sup>[32]</sup>      |           |   |
| Blogs <sup>[10]</sup>            | 5.8       | owl (5.80), whistle (5.81), humble (5.86)         |
| Dante's Inferno <sup>[?]</sup>   | 5.5       | glacier (5.50), repentant (5.53), mischief (5.57) |
| Heavy Metal                      | 5.4       | lamp (5.41), elevator (5.44), truck (5.47)        |
| lyrics <sup>[10]</sup>           |           |   |
|                                  |           |   |

# Lexicon Valley, Episode #62, June 17, 2015

# LEX I CON VALLEY

# Slate



Language has a Positivity Bias. How did we measure that?

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Measurement

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### Data sets:

#### Blog phrases containing "I feel...", "I am feeling", etc., taken from wefeelfine.org C (API, 2005–2010)

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An exploration of human emotion. in six movements





#### Jonathan Harris:

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🙈 2007 TED Talk: The Web's secret stories 🗹

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WN S

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Introduction

So I really consider myself a storyteller. But I don't really tell stories in the usual way, in the sense that I don't usually tell my own stories. Instead, I'm really interested in building tools that allow large numbers of other people to tell their stories, people all around the world. I do this because I think that people actually have a lot in common. I think people are very similar, but I also think that we have trouble seeing that. Measurement Happiness Some motions content Hedonometer Analysis Songs Blogs SOTU Geography Movement Other Emotions

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# wefeelfine.org:

 i feel very lonely and unoticed and that i am poised in a point of my life when i am able to do great things but just cant quite get them started

March 30, 2006 / from a 31 year old in fairfax virginia united states when it was cloud

- i feel lonely recently March 30, 2006 / from someone in georgia united state
- i feel lonely things are all good but i miss the way things used to be March 33, 2006 / from an 18 year old female in arizona united states
- i feel really lonely every night because i dont have any good friends irl that i can just talk about anything with March 31, 2006 / from a 17 year old male in lawrenceville georgia united states
- i feel really lonely and like any sensible loser i have to write about it in a blog March 31, 2006 / from an 18 year old male in missouri united states
- i feel so lonely inside March 31, 2005 / from a 24 year old male in san diego california united states when it was cloudy
- i feel soooooo lonely sometimes March 31, 2006 / from a 19 year old female in ellensburg washington united states
- i feel lonely March 31, 2006 / from someone
- i feel lonely i feel scared March 31, 2006 / from a 29 year old in mount version ohlo united states



i feel lonely when im with her March 31, 2006 / from someone in florida united state COcoNuTS @networksvox

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# wefeelfine.org:

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Happiness

| Feeling lovesick   | Gender Female | Age 20 - 29  | Weather Cloudy | Location All   | Date Feb 14, 2006   |  |
|--|---------------|--|----------------|--|---|--|
| All Feelings<br>100SEI   | Both Genders  | All Ages   | All Weather    | All Locations  | All Dates   |  |
| A looset<br>logacious<br>logacious<br>lost<br>lost<br>lost<br>lovable<br>lovable<br>lovable<br>lovels<br>loved<br>lovels<br>loved<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverl<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverly<br>loverloverly<br>loverly<br>loverly<br>loverly<br>loverloverloverly<br>loverly<br>loverlover 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darussalam<br>bulgaria<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>camada<br>ca 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| Tetl<br>Field  |               |  |                |  | d Feelings  |  |
|  |               |  |                | The second second  |   |  |

# wefeelfine.org:

FEEL

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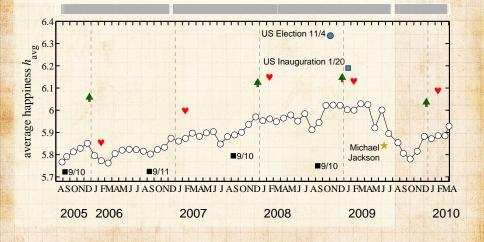
i feel so joyous at Aug 29, 2006 / from a female in wa



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## Blogs—Overall trend



#### From wefeelfine.org C by Jonathan Harris & Sep Kamvar



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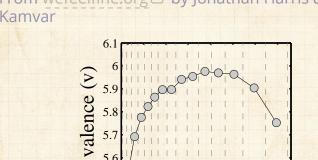
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Average happiness as a function of the age bloggers report they will turn in the year of their posting.

50 60

blogger age

70 80



5.6

5.5

13 20

30

## Blogs—Age

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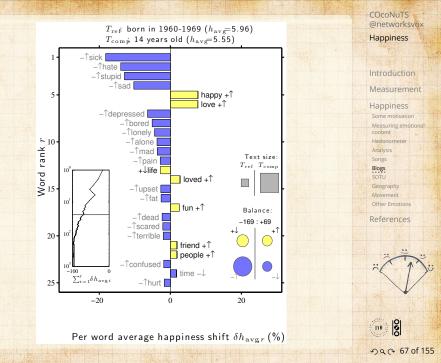


Self-report studies find little variation in happiness with age <sup>[12, 13]</sup>

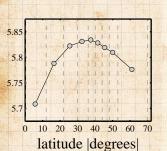
- Surprising: Expect a rise and fall.
- 🚳 A 'challenge' for theory...
- Related to the Easterlin Paradox: Money doesn't buy happiness
- But maybe it does a little bit—Veenhoven & Hagerty (2003) and Wolfers & Stevenson (2008).

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# Blogs—Latitude



#### Near equator—social factors

 Increase in 'sad', 'bored', 'lonely', 'stupid', 'guilty'
 Decrease in 'good' and 'people'

Near poles social/psychological/climate

- Increase in 'sick', 'guilty', 'cold', 'depressed', and 'headache' and decrease of 'love' and 'life.'
- Offset by decrease in 'hurt' and 'pain.'
- More 'bed' and 'sleep.'

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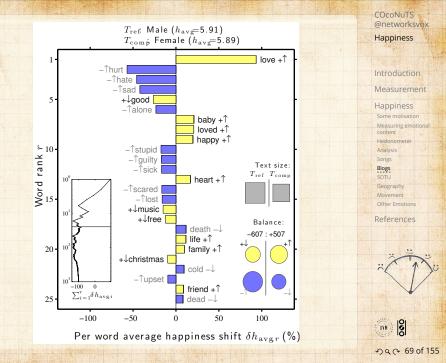
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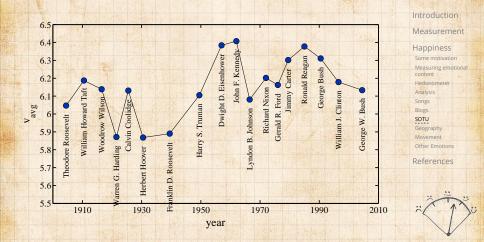


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## Presidential happiness:

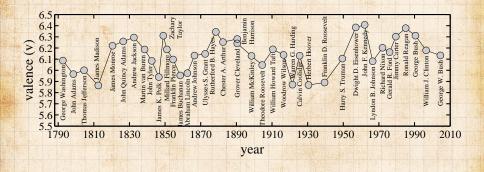
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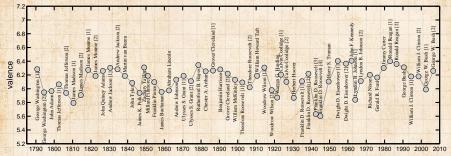


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# Presidential happiness:





year

# labMT 1.0: language assessment by Mechanical Turk

Make Money

TTs - Numan Intelligence Tasks - are individual tasks that you work an, find #The new

hanised Turk Marker ve

amazonmechanical turk

Nordan CTD: News Dealers &

Mechanical Turk is a marketplace for work

261,700 HITs available. View them m

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5000 most frequently used words for each corpus.
 10,222 words, 50 evaluations each, 1–9 scale: <sup>[27]</sup>

5



Four corpora:

**Google Books** 

**Music Lyrics** 

🔧 Twitter

#### PLOS ONE 10 Year Anniversary: Datasets

By sharing their underlying data and related metadata, authors increase the value and impact of their research. Since its incoption, PLOS OVE has encouraged data sharing, and in 2014, in keeping with our commitment to openness, reproducibility and scientific progress, the journal began requiring authors to make their data publick available

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Temporal Patterns of Happiness and Information in a Global Social Network: Hedonometrics and Twitter Peter Sheridan Dodds, Kameron Decker Harris, Isabel M. Kloumann, Catherine A. Blisa, Christopher M. Danforth

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Insight into this selection by Sune Lehmann "This paper has made an MTurk generated list of wordvalences openly available to the research community. As a ...

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|---------------------|-------------------|--------------|--------------|-----------------|-----------------|-------------|----------------|-----------------------------|
| <br>valence<br>rank | word              | valence      | std dev      | twitter<br>rank | g-books<br>rank | nyt<br>rank | lyrics<br>rank | @networksvo<br>Happiness    |
|                     |                   |              |              |                 |                 |             |                |                             |
| 1                   | laughter          | 8.50         | 0.93         | 3600            | -               | -           | 1728           |                             |
| 2<br>3              | happiness<br>love | 8.44<br>8.42 | 0.97         | 1853<br>25      | 2458<br>317     | 328         | 1230<br>23     | Introduction                |
| 3<br>4              |                   | 8.42         | 0.99         | 25<br>65        | 1372            | 1313        | 23<br>375      |                             |
| 4                   | happy<br>laughed  | 8.30         | 1.16         | 3334            | 3542            | 1313        | 2332           | Measuremen                  |
| 5<br>6<br>7         | laugh             | 8.20         | 1.37         | 1002            | 3998            | 4488        | 647            | Happiness                   |
| 7                   | laughing          | 8.20         | 1.11         | 1579            | -               |             | 1122           | Some motivation             |
| 8                   | excellent         | 8.18         | 1.10         | 1496            | 1756            | 3155        |                |                             |
| 9                   | laughs            | 8.18         | 1.16         | 3554            | -               | -           | 2856           | Measuring emotio<br>content |
| 10                  | joy               | 8.16         | 1.06         | 988             | 2336            | 2723        | 809            | Hedonometer                 |
| 11                  | successful        | 8.16         | 1.08         | 2176            | 1198            | 1565        | -              | Analysis                    |
| 12                  | win               | 8.12         | 1.08         | 154             | 3031            | 776         | 694            | Songs                       |
| 13                  | rainbow           | 8.10         | 0.99         | 2726            | _               | the has     | 1723           | Blogs                       |
| 14                  | smile             | 8.10         | 1.02         | 925             | 2666            | 2898        | 349            | SOTU                        |
| 15                  | won               | 8.10         | 1.22         | 810             | 1167            | 439         | 1493           | Geography                   |
| 16                  | pleasure          | 8.08         | 0.97         | 1497            | 1526            | 4253        | 1398           | Movement                    |
| 17                  | smiled            | 8.08         | 1.07         | -               | 3537            | -           | 2248           | Other Emotions              |
| 18                  | rainbows          | 8.06         | 1.36         | 12 4 Star       | -               | -           | 4216           | References                  |
| 19                  | winning           | 8.04         | 1.05         | 1876            | -               | 1426        | 3646           | References                  |
| 20                  | celebration       | 8.02         | 1.53         | 3306            | -               | 2762        | 4070           |                             |
| 21                  | enjoyed           | 8.02         | 1.53         | 1530            | 2908            | 3502        | -              |                             |
| 22                  | healthy           | 8.02         | 1.06         | 1393            | 3200            | 3292        | 4619           |                             |
| 23                  | music             | 8.02         | 1.12         | 132             | 875             | 167         | 374            |                             |
| 24                  | celebrating       | 8.00         | 1.14         | 2550            |                 | -           | -              | 1                           |
| 25                  | congratulations   | 8.00         | 1.63         | 2246            |                 | -           |                |                             |
| 26                  | weekend           | 8.00         | 1.29         | 317             | -               | 833         | 2256           |                             |
| 27                  | celebrate         | 7.98         | 1.15         | 1606            | -               | 3574        | 2108           |                             |
| 28                  | comedy            | 7.98         | 1.15         | 1444            | -               | 2566        | -              |                             |
| 29                  | jokes             | 7.98         | 0.98         | 2812            | -               | -           | 3808           | 101                         |
| 30                  | rich              | 7.98         | 1.32         | 1625            | 1221            | 1469        | 890            |                             |
| 0                   | 0                 | 0            | 0            | 0               | 0               | 0           | 0              |                             |
|                     |                   |              |              |                 |                 |             |                |                             |

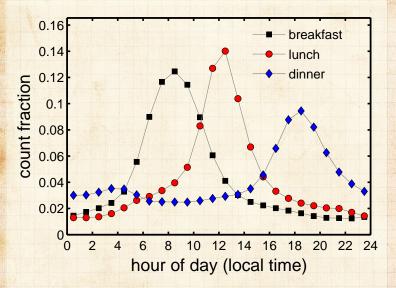
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COcoNuTS @networksvc

|                 |           |         |         |                 |                     |             |                | COcoNuTS                  |
|-----------------|-----------|---------|---------|-----------------|---------------------|-------------|----------------|---------------------------|
| valence<br>rank | word      | valence | std dev | twitter<br>rank | g-books<br>rank     | nyt<br>rank | lyrics<br>rank | @networksvox<br>Happiness |
|                 |           | A SIL   |         |                 | C. Mathada          |             | have the state |                           |
| 0               | 0         | 0       |         | 0               |                     | 0           | 0              |                           |
| 10193           | violence  | 1.86    | 1.05    | 4299            | 1724                | 1238        | 2016           | Introduction              |
| 10194           | cruel     | 1.84    | 1.15    | 2963            | 30 <del>1</del>     | -           | 1447           | introduction              |
| 10195           | cry       | 1.84    | 1.28    | 1028            | 3075                |             | 226            | Measurement               |
| 10196           | failed    | 1.84    | 1.00    | 2645            | 1618                | 1276        | 2920           |                           |
| 10197           | sickness  | 1.84    | 1.18    | 4735            | -                   | -           | 3782           | Happiness                 |
| 10198           | abused    | 1.83    | 1.31    | 1               | The last            | -           | 4589           | Some motivation           |
| 10199           | tortured  | 1.82    | 1.42    | -               |                     |             | 4693           | Measuring emotional       |
| 10200           | fatal     | 1.80    | 1.53    | -               | 4089                | -           | 3724           | content                   |
| 10201           | killings  | 1.80    | 1.54    | -               | -                   | 4914        | 2 - 12         | Hedonometer               |
| 10202           | murdered  | 1.80    | 1.63    | -               |                     |             | 4796           | Analysis                  |
| 10203           | war       | 1.80    | 1.41    | 468             | 175                 | 291         | 462            | Songs                     |
| 10204           | kills     | 1.78    | 1.23    | 2459            | - 51                | -           | 2857           | Blogs                     |
| 10205           | jail      | 1.76    | 1.02    | 1642            | -                   | 2573        | 1619           | SOTU                      |
| 10206           | terror    | 1.76    | 1.00    | 4625            | 4117                | 4048        | 2370           | Geography                 |
| 10207           | die       | 1.74    | 1.19    | 418             | 730                 | 2605        | 143            | Movement                  |
| 10208           | killing   | 1.70    | 1.36    | 1507            | 4428                | 1672        | 998            | Other Emotions            |
| 10209           | arrested  | 1.64    | 1.01    | 2435            | 4474                | 1435        | -              | References                |
| 10210           | deaths    | 1.64    | 1.14    | -               | -                   | 2974        |                | nererences                |
| 10211           | raped     | 1.64    | 1.43    | -               | -                   | -           | 4528           |                           |
| 10212           | torture   | 1.58    | 1.05    | 3175            | -                   | E State     | 3126           |                           |
| 10213           | died      | 1.56    | 1.20    | 1223            | 866                 | 208         | 826            | 00                        |
| 10214           | kill      | 1.56    | 1.05    | 798             | 2727                | 2572        | 430            |                           |
| 10215           | killed    | 1.56    | 1.23    | 1137            | 1603                | 814         | 1273           |                           |
| 10216           | cancer    | 1.54    | 1.07    | 946             | 1884                | 796         | 3802           |                           |
| 10217           | death     | 1.54    | 1.28    | 509             | 307                 | 373         | 433            |                           |
| 10218           | murder    | 1.48    | 1.01    | 2762            | 3110                | 1541        | 1059           |                           |
| 10219           | terrorism | 1.48    | 0.91    | -               | 4                   | 3192        | -              |                           |
| 10220           | rape      | 1.44    | 0.79    | 3133            | 11 <u>1</u> 1111111 | 4115        | 2977           |                           |
| 10221           | suicide   | 1.30    | 0.84    | 2124            | 4707                | 3319        | 2107           |                           |
| 10222           | terrorist | 1.30    | 0.91    | 3576            | _                   | 3026        | -              | iol the second second     |
|                 |           |         |         |                 |                     |             |                |                           |

| std dev<br>rank | word           | valence | std dev | twitter<br>rank | g-books<br>rank | nyt<br>rank   | lyrics | COcoNuTS<br>@networksvox<br>Happiness   |
|-----------------|----------------|---------|---------|-----------------|-----------------|---------------|--------|---|
| TUTIK           |                |         |         | Тапк            | TUTIK           | TUTIK         | TUTIK  | nappiness   |
| 1               | f مر مر الم    | 4.64    | 2.93    | 448             | -               | _             | 620    |   |
| 2               | f**kin         | 3.86    | 2.74    | 1077            |                 | I             | 688    |   |
| 3               | f**ked         | 3.56    | 2.74    | 1840            | _               | 1             | 904    | Introduction  |
| 4               | pussy          | 4.80    | 2.66    | 2019            |                 | 1             | 949    |   |
| 5               | whiskey        | 5.72    | 2.64    |                 |                 |               | 2208   | Measurement   |
| 6               | slut           | 3.57    | 2.63    |                 | -               | -             | 4071   | Happiness   |
| 7               |                | 3.37    | 2.63    | 1201            |                 | T             | 3279   | a second s |
| 8               | cigarettes     | 4.14    |         | - 322           |                 | T             | 185    | Some motivation   |
|                 | f☆☆k           |         | 2.58    | 322             | -               | -             | 185    | Measuring emotiona<br>content   |
| 9               | mortality      | 4.38    | 2.55    | 1               | 3960            | Tank          | -      | Hedonometer   |
| 10              | cigarette      | 3.09    | 2.52    | 1               | 7               | T             | 2678   | Analysis  |
| 11              | motherf★★kers  | 2.51    | 2.47    | -               | -               | + 1           | 1466   | Songs   |
| 12              | churches       | 5.70    | 2.46    | T               | 2281            | -             | -      | Blogs   |
| 13              | motherf★★king  | 2.64    | 2.46    | -               | - 81            | -             | 2910   | SOTU  |
| 14              | capitalism     | 5.16    | 2.45    | -               | 4648            | -             | -      |   |
| 15              | porn           | 4.18    | 2.43    | 1801            | -               | /1- 1 4 4 4 H | - 124  | Geography   |
| 16              | summer         | 6.40    | 2.39    | 896             | 1226            | 721           | 590    | Movement  |
| 17              | beer           | 5.92    | 2.39    | 839             | 4924            | 3960          | 1413   | Other Emotions  |
| 18              | execution      | 3.10    | 2.39    | 12-5            | 2975            | -             | -      | References  |
| 19              | wines          | 6.28    | 2.37    | 1               | _               | 3316          | -      | nererenees  |
| 20              | zombies        | 4.00    | 2.37    | 4708            | _               | 101           | 12000  |   |
| 21              | aids           | 4.28    | 2.35    | 2983            | 3996            | 1197          |        |   |
| 22              | capitalist     | 4.84    | 2.34    | _               | 4694            | _             |        |   |
| 23              | revenge        | 3.71    | 2.34    |                 | _               | 1             | 2766   |   |
| 24              | mcdonalds      | 5.98    | 2.33    | 3831            |                 |               | -      | 5   |
| 25              | beatles        | 6.44    | 2.33    | 3797            |                 |               |        |   |
| 26              | islam          | 4.68    | 2.33    | 5151            | 4514            | 1             |        |   |
| 27              |                | 5.30    | 2.33    | 627             | 769             | 460           | 499    |   |
| 27              | pay<br>alcohol | 5.20    | 2.32    | 2787            | 2617            | 3752          | 3600   |   |
| 28<br>29        | muthaf**kin    | 3.00    | 2.32    | 2/8/            | 2017            | 3/52          | 4107   |   |
|                 |                |         |         | -               | -               | 4000          |        | 19  |
| 30              | christ         | 6.16    | 2.31    | 2509            | 909             | 4238          | 1526   | UNN S   |
| 0               | 0              | 0       | 0       |                 | 0               | 0             | 0      |   |

### Twitter—living in the now:



🚳 Quantifying the quotidian.

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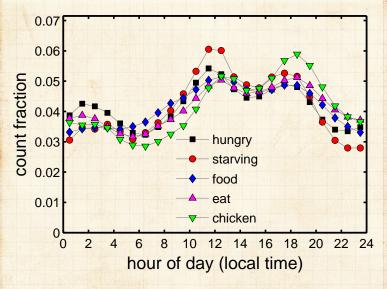
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Makes the unexpected believable...

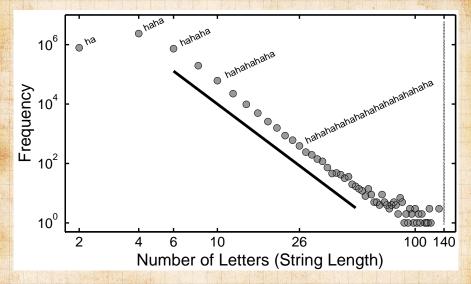
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# The happiest distribution:



Work by Tyler Gray et al., unpublished.



#### Tom Sherwood @tomsherwood 1 Oct Mississippi congressman moves gate and lets 92 WW2 vets see their monument. pic.twitter.com/541GvLLITz

Retweeted 442 times



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# Breaking: Two Explosions in the White House and Barack Obama is injured



10:07 AM - 23 Apr 13

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# Mentions of CIA on Twitter, end of 2011:



COcoNuTS @networksvox Happiness Introduction Measurement Happiness

content Hedonometer Analysis Songs Blogs SOTU Geography Movement Other Emotions

References

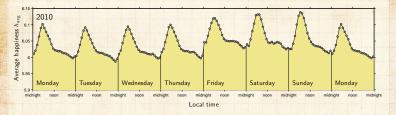
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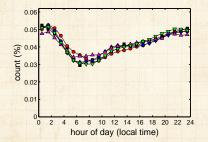
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See story here ☑ for example [slate].

# The daily unravelling of the human mind:





COCONUTS @networksvox Happiness

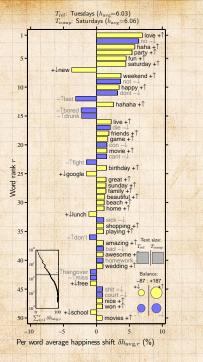
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References

Social Scientists wade into the Tweet stream" by Greg Miller,



UVN





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Introduction

Text element and context correlate in happiness scores:

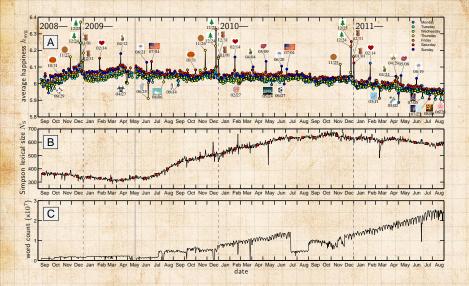
- Compare ambient happiness with text element happiness.
- Spearman correlation coefficient:  $r_{e} \simeq 0.79$ , *p*-value  $< 10^{-10}$ .
- An on-average result: says nothing about any individual sentence.
- Extra random piece: stemming is fallible.

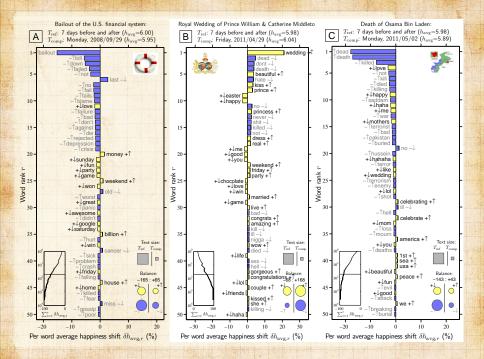
Measurement
Happiness
Some motivation
Measuring emotional
content
Hedonometer
Analysis
Songs
Bilogs
SOTU
Geography
Movement
Other Emotions

References



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# Ambient happiness:

| Word                             | $h_{\rm avg}^{({\rm amb})}$ | Total Tweets        | h <sup>(norm)</sup>      | Word                | $h_{\rm avg}^{({\rm amb})}$ | Total Tweets       | h <sup>(norm)</sup> |
|----------------------------------|-----------------------------|---------------------|--------------------------|---------------------|-----------------------------|--------------------|---------------------|
| 1. happy                         | +0.430                      | 1.65e+07(13)        | +1.104(1)                | 51. snow            | -0.051                      | 2.60e+06 (49)      | +0.083(39)          |
| 2. Christmas                     | +0.404                      | 4.89e+06 (35)       | +0.953(3)                | 52. Jon Stewart     | -0.052                      | 5.21e+04 (97)      | -0.024 (48)         |
| 3. vegan                         | +0.315                      | 1.84e + 05(90)      | -0.015 (46)              | 53. school          | -0.056                      | 9.26e+06 (24)      | +0.050(42)          |
| 4. :)                            | +0.274                      | 1.04e + 07(20)      | +0.630(12)               | 54. Lehman Brothers | -0.078                      | 8.50e+03 (100)     | -0.721 (79)         |
| 5. family                        | +0.251                      | 5.01e+06 (32)       | +0.716(7)                | 55. them            | -0.090                      | 1.54e+07 (15)      | -0.280 (60)         |
| 6. :-)                           | +0.228                      | 1.67e + 06(60)      | +0.560(16)               | 56. right           | -0.090                      | 1.92e+07 (10)      | +0.126(35)          |
| 7. our                           | +0.207                      | 1.41e+07 (16)       | +0.159(33)               | 57. woman           | -0.115                      | 2.54e+06 (51)      | +0.202(30)          |
| 8. win                           | +0.204                      | 7.98e+06 (26)       | +0.924(4)                | 58. left            | -0.118                      | 4.89e+06 (34)      | -0.383 (63)         |
| 9. vacation                      | +0.200                      | 9.35e+05 (67)       | +0.817(5)                | 59. me              | -0.119                      | 1.44e + 08(4)      | +0.160(32)          |
| 10. party                        | +0.170                      | 6.44e + 06(29)      | +0.679(9)                | 60. election        | -0.127                      | 5.60e+05 (75)      | -0.306 (61)         |
| 11. love                         | +0.164                      | 4.67e+07 (6)        | +0.977(2)                | 61. Sarah Palin     | -0.128                      | 2.26e+05 (87)      | -0.681 (76)         |
| 12. friends                      | +0.155                      | 7.67e+06 (27)       | +0.685(8)                | 62. no              | -0.132                      | 9.51e+07 (5)       | -1.415 (90)         |
| 13. hope                         | +0.149                      | 1.18e + 07(18)      | +0.515(19)               | 63. rain            | -0.134                      | 3.23e+06(41)       | +0.050(44)          |
| 14. coffee                       | +0.147                      | $2.80e \pm 06(46)$  | +0.518(18)               | 64. climate         | -0.135                      | 3.64e+05 (80)      | -0.160 (51)         |
| 15. cash                         | +0.146                      | 1.28e + 06(63)      | +0.601(14)               | 65. gav             | -0.152                      | 2.73e+06 (47)      | -0.552 (72)         |
| 16. sun                          | +0.144                      | 2.39e + 06(52)      | +0.737(6)                | 66. lose            | -0.157                      | 2.06e+06 (55)      | -1.181 (86)         |
| 17. income                       | +0.137                      | 5.10e+05 (76)       | +0.621(13)               | 67. they            |                             | 2.74e+07 (8)       | -0.208 (58)         |
| 18. summer                       | +0.135                      | 3.00e + 06(43)      | +0.221(29)               | 68. oil             | -0.162                      | 1.38e + 06(62)     | -0.411 (65)         |
| 19. church                       | +0.131                      | 1.81e + 06(58)      |                          | 69. cold            | -0.162                      | 3.67e+06 (36)      | -0.546 (71)         |
| 20. Valentine                    | +0.127                      | 2.47e+05 (84)       | $\pm 0.593(15)$          | 70. I feel          |                             | 5.17e+06 (31)      | -0.129 (50)         |
| 21. Stephen Colbert              | $\pm 0.126$                 | $2.38e \pm 04$ (99) | +0.001(45)               | 71 man              | -0.175                      | $1.59e \pm 07(14)$ | -0.163 (52)         |
| 22. USA                          |                             | 2.16e + 06(54)      | +0.325(26)               | 72. Republican      |                             | 2.30e + 05(86)     | -0.539 (70)         |
| 23 1                             |                             | 3.44e+06 (40)       |                          |                     |                             | 3.56e+06 (38)      | -1.366 (89)         |
| 24 winter                        | +0.101                      | 1.26e + 06(64)      | +0.050(43)               | 74. gas             |                             | $1.02e \pm 06(65)$ | -0.471 (67)         |
| 25 God                           | +0.099                      | 8.58e+06 (25)       | +0.468(20)               | 75. economy         |                             | 6.09e+05 (73)      | -0.525 (69)         |
| 26 hot                           |                             | 7.12e + 06(28)      | -0.172 (54)              | 76. Obama           |                             | $2.98e \pm 06(44)$ | -0.173 (55)         |
| 27. ;)                           |                             | 2.61e+06 (48)       | +0.326(25)               | 77 Democrat         |                             | $9.32e \pm 04(93)$ | -0.384 (64)         |
| 28 Jesus                         |                             |                     |                          | 78. Congress        |                             | $3.92e \pm 05(79)$ | -0.580 (74)         |
| 29. today                        |                             | 2.56e+07 (9)        | +0.126(36)               |                     |                             | 6.27e+06 (30)      | -1.551 (96)         |
| 30. kiss                         |                             | $1.70e \pm 06(59)$  | +0.632(11)               | 80. sick            |                             | $3.58e \pm 06(37)$ | -1.630 (97)         |
| 31. ves                          |                             | $1.16e \pm 07(19)$  | +0.321(27)               |                     |                             | 2.15e+05(88)       | -0.569 (73)         |
| 32. tomorrow                     | +0.054                      | 1.04e+07(21)        | +0.086(38)               | 82. war             | -0.270                      | 1.96e+06(57)       | -2.040 (100)        |
| 33. vou                          |                             | 1.73e+08(3)         | +0.111(37)               |                     |                             | 1.52e + 05(91)     | -0.316 (62)         |
| 34. heaven                       | +0.032<br>+0.041            | 7.42e+05(71)        | +0.674(10)               | 84 hate             |                             | 9.65e+06(23)       | -1.520 (94)         |
| 35. ;-)                          |                             | 9.39e+05 (66)       | +0.395(23)               | 85. Glenn Beck      |                             | 1.14e+05 (92)      | -0.776 (82)         |
| 36. we                           |                             | 3.91e+07(7)         | +0.335(23)<br>+0.146(34) |                     |                             | 1.87e+05(82)       | -0.710 (78)         |
| 37. vesterday                    |                             | 3.08e+06(42)        | -0.168 (53)              | 87. George Bush     |                             | 3.23e+04(98)       | -0.747 (80)         |
| 38. dark                         |                             | $1.58e \pm 06(42)$  | -0.766 (81)              | 88. Goldman Sachs   |                             | 5.27e+04 (96)      | -0.984 (84)         |
| 39. ?                            |                             | 2.32e+06(53)        | -0.503 (68)              | 89. depressed       | -0.339                      | 2.81e+05 (82)      | -1.541 (95)         |
| 40 RT                            |                             | $3.39e \pm 08(1)$   | -0.443 (66)              | 90. Senate          |                             | $4.48e \pm 05(78)$ | -0.601 (75)         |
| 41. Michael Jackson              |                             | 8.26e+05 (70)       | -0.443 (00)              | 91. BP              |                             | 5.82e+05 (74)      | -0.902 (83)         |
| 41. Michael Jackson<br>42. night |                             |                     |                          | 91. BP<br>92. gun   | -0.355                      |                    | -0.902 (83)         |
| 42. night<br>43. life            |                             |                     |                          |                     |                             |                    |                     |
| 43. me<br>44. health             |                             |                     | +0.422(22)               |                     |                             | 5.10e+05 (77)      | -1.452 (91)         |
|                                  | -0.000                      | 2.58e+06 (50)       | +0.447(21)               | 94. headache        | -0.437                      | 8.57e+05 (69)      | -1.881 (98)         |
| 45. sex                          |                             | 3.55e+06(39)        | +0.542(17)               |                     |                             | 3.40e+05(81)       | -1.174 (85)         |
| 46. work                         | -0.010                      | 1.84e + 07(11)      | -0.174(56)               | 96. :(              |                             | 2.89e + 06(45)     | -1.288 (88)         |
| 47. girl                         | -0.010                      | 1.01e+07(22)        |                          | 97. Afghanistan     |                             | 2.74e+05(83)       | -1.458 (92)         |
| 48. boy                          | -0.026                      | 4.93e+06(33)        |                          | 98. mosque          |                             | 6.98e+04 (95)      | -0.694 (77)         |
| 49. I                            | -0.048                      | 3.08e + 08(2)       | -0.062 (49)              | 99. flu             |                             | 9.01e+05(68)       | -1.912 (99)         |
| 50. commute                      | -0.048                      | $9.01e \pm 04$ (94) | -0.206(57)               | 100. Iraq           | -0.773                      | 2.39e+05 (85)      | -1.282 (87)         |

#### COcoNuTS @networksvox

Happiness

#### Introduction

Measurement

- Happiness
- Some motivation
- Measuring emotional content
- Hedonometer
- Analysis
- Songs
- Blogs
- SOTU
- Geography Movement
- Other Emotions

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UVM 000

# Simpson lexical size, $N_{\rm S}$ :

| Word                 | $h_{\rm avg}^{({\rm amb})}$ | Total Tweets                               | h <sup>(norm)</sup> <sub>avg</sub> | Word                          | $h_{\rm avg}^{({\rm amb})}$ | Total Tweets                 | h <sup>(norm)</sup> |
|----------------------|-----------------------------|--|------------------------------------|-------------------------------|-----------------------------|------------------------------|---------------------|
| 1. happy             | +0.430                      | 1.65e+07(13)                               | +1.104(1)                          | 51. snow                      | -0.051                      | 2.60e+06 (49)                | +0.083(39)          |
| 2. Christmas         | +0.404                      | 4.89e+06 (35)                              | +0.953(3)                          | 52. Jon Stewart               | -0.052                      | 5.21e+04 (97)                | -0.024 (48)         |
| 3. vegan             |                             | 1.84e+05 (90)                              | -0.015 (46)                        | 53. school                    |                             | 9.26e+06 (24)                | +0.050(42)          |
| 4. :)                | +0.274                      | 1.04e + 07(20)                             | +0.630(12)                         |                               | -0.078                      | 8.50e+03 (100)               | -0.721 (79)         |
| 5. family            | +0.251                      | 5.01e+06 (32)                              | +0.716(7)                          | 55. them                      | -0.090                      | 1.54e+07 (15)                | -0.280 (60)         |
| 6. :-)               | +0.228                      | 1.67e+06 (60)                              | +0.560(16)                         | 56. right                     | -0.090                      | 1.92e+07 (10)                | +0.126(35)          |
| 7. our               | +0.207                      | 1.41e+07 (16)                              | +0.159(33)                         | 57. woman                     | -0.115                      | 2.54e+06 (51)                | +0.202(30)          |
| 8. win               | +0.204                      | 7.98e+06 (26)                              | +0.924(4)                          | 58. left                      | -0.118                      | 4.89e+06 (34)                | -0.383 (63)         |
| 9. vacation          | +0.200                      | 9.35e+05 (67)                              | +0.817(5)                          | 59. me                        | -0.119                      | 1.44e+08 (4)                 | +0.160(32)          |
| 10. party            | +0.170                      | 6.44e+06 (29)                              | +0.679(9)                          | 60. election                  | -0.127                      | 5.60e+05 (75)                | -0.306 (61)         |
| 11. love             | +0.164                      | 4.67e+07 (6)                               | +0.977(2)                          | 61. Sarah Palin               | -0.128                      | 2.26e+05 (87)                | -0.681 (76)         |
| 12. friends          | +0.155                      | 7.67e+06 (27)                              | +0.685(8)                          | 62. no                        | -0.132                      | 9.51e+07 (5)                 | -1.415 (90)         |
| 13. hope             | +0.149                      | 1.18e + 07(18)                             | +0.515(19)                         | 63. rain                      | -0.134                      | 3.23e+06 (41)                | +0.050(44)          |
| 14. coffee           | +0.147                      | 2.80e + 06(46)                             | +0.518(18)                         | 64. climate                   | -0.135                      | 3.64e+05 (80)                | -0.160 (51)         |
| 15. cash             | +0.146                      | 1.28e + 06(63)                             | +0.601(14)                         | 65. gay                       | -0.152                      | 2.73e+06 (47)                | -0.552 (72)         |
| 16. sun              | +0.144                      | 2.39e + 06(52)                             | +0.737(6)                          | 66. lose                      | -0.157                      | 2.06e+06 (55)                | -1.181 (86)         |
| 17. income           | +0.137                      | 5.10e+05 (76)                              | +0.621(13)                         | 67. they                      | -0.159                      | 2.74e+07 (8)                 | -0.208 (58)         |
| 18. summer           | +0.135                      | 3.00e + 06(43)                             | +0.221(29)                         | 68. oil                       |                             | 1.38e + 06(62)               | -0.411 (65)         |
| 19. church           | +0.131                      | $1.81e \pm 06(58)$                         | -0.016 (47)                        | 69. cold                      | -0.162                      | 3.67e+06 (36)                | -0.546 (71)         |
| 20. Valentine        | +0.127                      | 2.47e+05 (84)                              | +0.593(15)                         | 70. I feel                    | -0.173                      | 5.17e+06 (31)                | -0.129 (50)         |
| 21. Stephen Colbert  | +0.126                      | 2.38e + 04(99)                             | +0.001(45)                         | 71. man                       |                             | 1.59e+07 (14)                | -0.163 (52)         |
| 22 USA               |                             | $2.16e \pm 06(54)$                         | $\pm 0.325(26)$                    | 72 Republican                 |                             | 2.30e+05 (86)                | -0.539 (70)         |
| 23 1                 | +0.106                      | $3.44e \pm 06(40)$                         | +0.195(31)                         | 73. sad                       |                             | 3.56e+06 (38)                | -1.366 (89)         |
| 24 winter            |                             | 1.26e + 06(64)                             | +0.050(43)                         | 74. gas                       |                             | 1.02e+06(65)                 | -0.471 (67)         |
| 25 God               |                             | 8.58e+06 (25)                              | +0.468(20)                         | 75. economy                   |                             | 6.09e+05 (73)                | -0.525 (69)         |
| 26. hot              |                             | $7.12e \pm 06(28)$                         | -0.172 (54)                        | 76. Obama                     |                             | 2.98e+06 (44)                | -0.173 (55)         |
| 27. :)               |                             | 2.61e+06 (48)                              | +0.326(25)                         | 77. Democrat                  |                             | 9.32e+04 (93)                | -0.384 (64)         |
| 28 Jesus             |                             | $2.03e \pm 06(56)$                         | +0.247(28)                         | 78. Congress                  |                             | 3.92e+05 (79)                | -0.580 (74)         |
| 29. today            |                             | $2.56e \pm 07(9)$                          | +0.126(36)                         | 79, hell                      |                             | 6.27e+06 (30)                | -1.551 (96)         |
| 30. kiss             |                             | 1.70e+06 (59)                              | +0.632(11)                         | 80 sick                       |                             | 3.58e+06 (37)                | -1.630 (97)         |
| 31. ves              |                             | 1.16e+07 (19)                              | +0.321(27)                         | 81 Muslim                     |                             | 2.15e+05 (88)                | -0.569 (73)         |
| 32. tomorrow         |                             | 1.04e+07(21)                               | +0.086(38)                         | 82. war                       |                             | 1.96e+06(57)                 | -2.040 (100         |
| 33. you              |                             | 1.73e + 08(3)                              | +0.111(37)                         | 83. Pope                      |                             | 1.52e+05 (91)                | -0.316 (62)         |
| 34 heaven            |                             | $7.42e \pm 05.(71)$                        | +0.674(10)                         | 84 hate                       |                             | $9.65e \pm 06.(23)$          | -1.520 (94)         |
| 35. ;-)              |                             | 9.39e+05 (66)                              | +0.395(23)                         | 85. Glenn Beck                |                             | 1.14e+05 (92)                | -0.776 (82)         |
| 36, we               |                             | 3.91e+07(7)                                | +0.146(34)                         | 86. Islam                     |                             | 1.87e+05 (89)                | -0.710 (78)         |
| 37. vesterday        |                             | 3.08e+06 (42)                              | -0.168 (53)                        | 87. George Bush               |                             | 3.23e+04 (98)                | -0.747 (80)         |
| 38. dark             |                             | $1.58e \pm 06(61)$                         | -0.766 (81)                        | 88. Goldman Sachs             |                             | 5.27e+04 (96)                | -0.984 (84)         |
| 39. ?                |                             | 2.32e + 06(53)                             | -0.503 (68)                        | 89. depressed                 |                             | 2.81e+05 (82)                | -1.541 (95)         |
| 40 BT                |                             | $3.39e \pm 08(1)$                          | -0.443 (66)                        | 90 Senate                     |                             | 4.48e+05(78)                 | -0.601 (75)         |
| 41. Michael Jackson  |                             | 8.26e + 05(70)                             | -0.213 (59)                        | 91 BP                         |                             | 5.82e+05 (74)                | -0.902 (83)         |
| 42. night            |                             | 1.71e+07(12)                               | +0.074(40)                         | 92. gun                       |                             | $6.81e \pm 05(72)$           | -1.476 (93)         |
| 43. life             |                             | 1.40e+07(17)                               |                                    | 93. drugs                     |                             | 5.10e+05 (72)                | -1.452 (91)         |
| 43. me<br>44. health | +0.012                      | 1.40e+07(17)<br>2.58e+06(50)               | +0.422(22)<br>+0.447(21)           | 93. drugs<br>94. headache     | -0.382                      | 5.10e+05(77)<br>8.57e+05(69) | -1.452 (91)         |
| 45. sex              |                             | 3.55e+06(30)                               |                                    | 95. :-(                       |                             | 3.40e+05 (81)                | -1.174 (85)         |
| 45. sex              | -0.008                      | $1.84e \pm 07(11)$                         | -0.174 (56)                        | 95. :-(<br>96. :(             |                             | $2.89e \pm 06(45)$           | -1.288 (88)         |
| 40. WORK<br>47. girl | -0.010                      | $1.84e \pm 07$ (11)<br>$1.01e \pm 07$ (22) |                                    | 90. :(<br>97. Afghanistan     |                             | 2.89e+06(45)<br>2.74e+05(83) | -1.288 (88)         |
| 47. giri<br>48. boy  | -0.010                      | 4.93e+06(33)                               |                                    | 97. Aignanistan<br>98. mosque |                             | 2.74e+05(83)<br>6.98e+04(95) | -0.694 (77)         |
| 48. boy<br>49 I      |                             |  |                                    |                               |                             |                              |                     |
| 49. 1<br>50. commute | -0.048                      | 3.08e + 08(2)                              | -0.062 (49)                        | 99. flu                       |                             | 9.01e+05 (68)                | -1.912 (99)         |
| ou. commute          | -0.048                      | $9.01e \pm 04$ (94)                        | -0.206 (57)                        | 100. Iraq                     | -0.773                      | 2.39e+05 (85)                | -1.282 (87)         |

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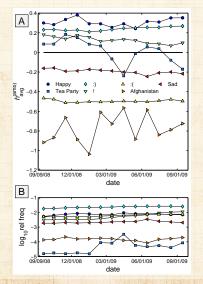
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# Ambient happiness:



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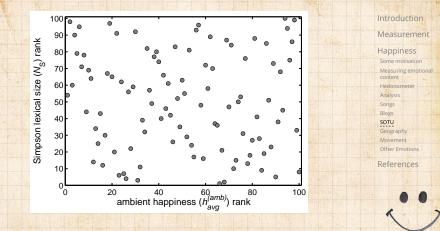
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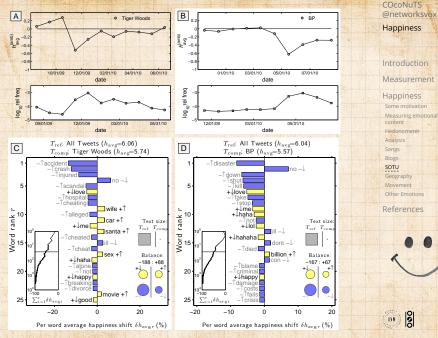
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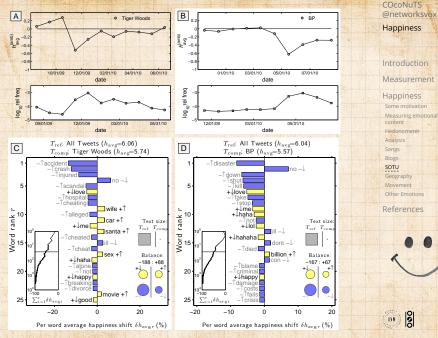
COcoNuTS @networksvox Happiness

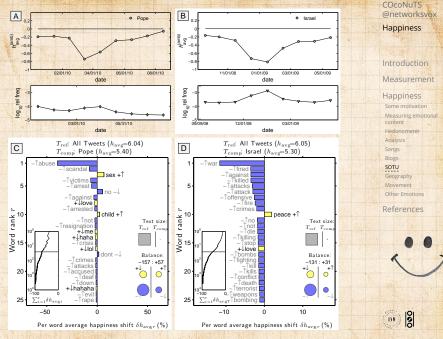


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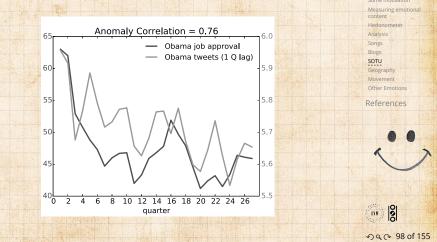


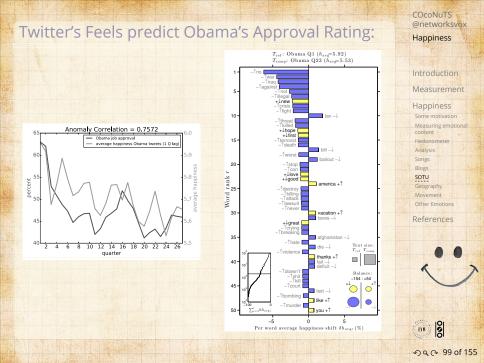
# "Public opinion polling with Twitter" Cody et al., Available online at https://arxiv.org/abs/1608.02024, 2016.<sup>[4]</sup>



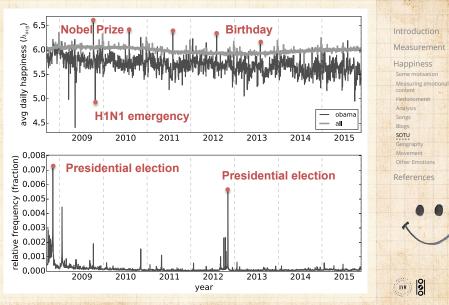
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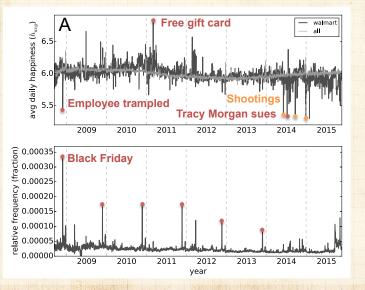


# Ambient happiness for "Obama":



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## Ambient happiness for "Walmart":



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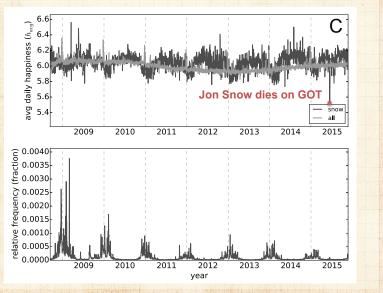
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# Ambient happiness for "snow":



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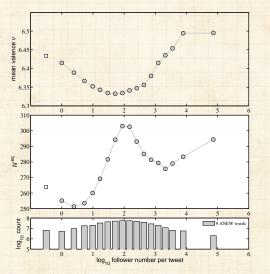
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## Dunbar number action:



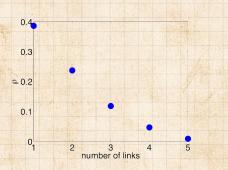
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Early Twitter data—follower counts are not worth 3 so much.

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UVN

"Twitter reciprocal reply networks exhibit assortativity with respect to happiness" Bliss, Kloumann, Harris, Danforth, and Dodds. Journal of Computational Science, **3**, 388–397, 2012.<sup>[1]</sup>



Decay in happiness correlation in social network.
 Not a test of contagion ...

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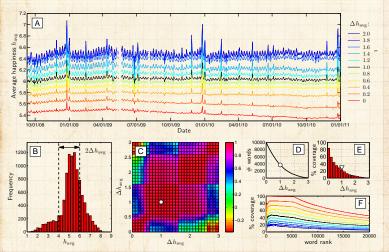
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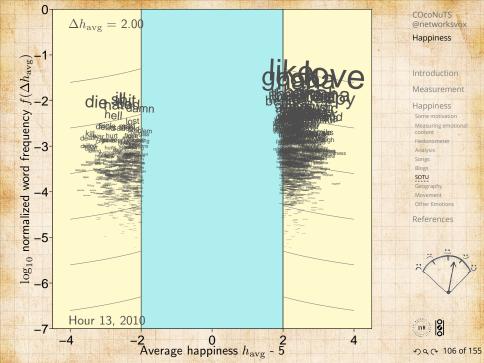
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# The very surprising tunable hedonometer:



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#### Online instrument: hedonometer.org

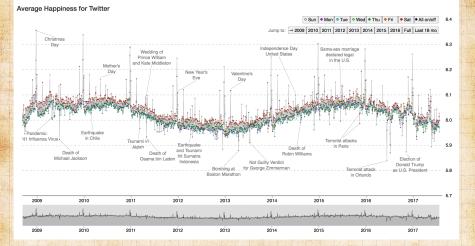
projects - About

Instructions Words

Blog Press Papers

ers Talks Fund us

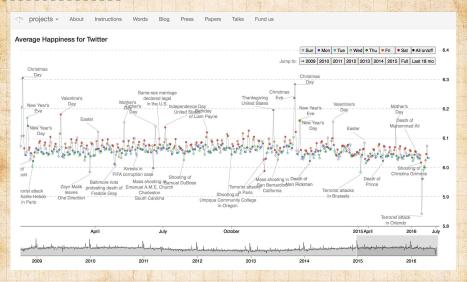
My Wordshifts ()



#### 🚳 Machine: @andyreagan 🗹

A Planned happiness versus tragedies.

hedonometer.org



## 🚳 Machine: @andyreagan 🖸

## hedonometer.org



🚳 Machine: @andyreagan 🗹

# The one reliable source of spontaneous, widespread happiness and despair:

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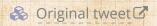
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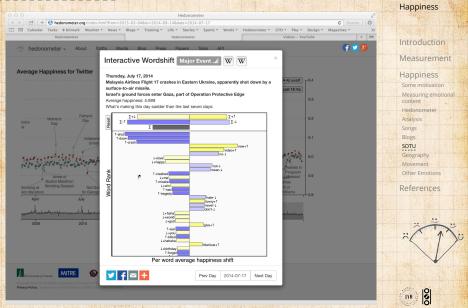
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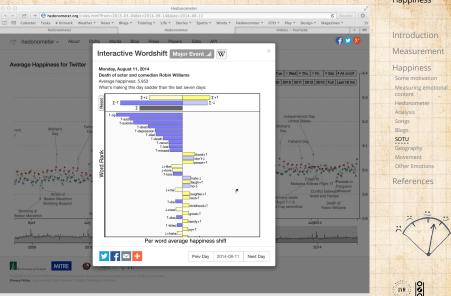




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COCONUTS

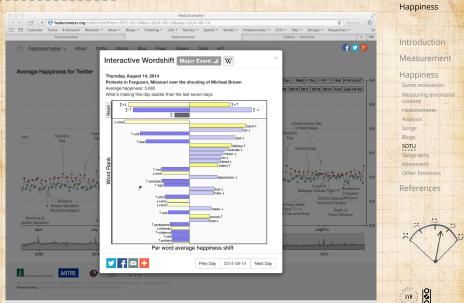
@networksvox



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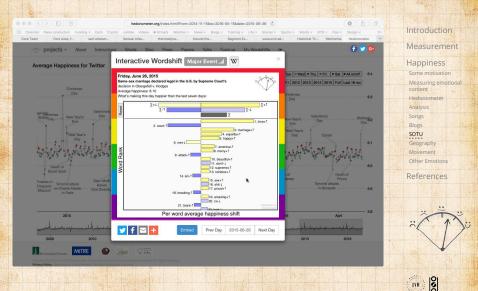
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COcoNuTS @networksvox

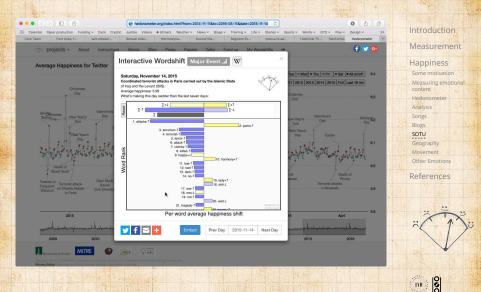
Happiness



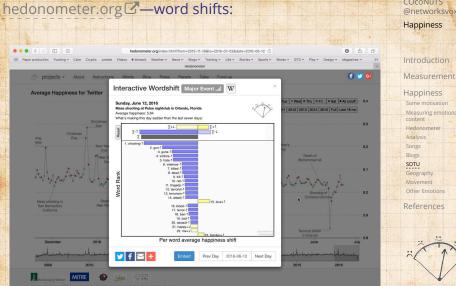
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@networksvox

| Makow    | our own:  |   |                         | COcoNuTS<br>@networksvox               |  |
|----------|---|---|-------------------------|--|--|
| iviake y | Happiness   |   |                         |  |  |
|          |   | hedenometer.org/wordshifterator/  | Main .                  |  |  |
|          | Paper production Funding Cryptic NYT Crossword Videos + bitmark           | Weather V News V Blogs V Training V Life V Stories V Sports V Words V G | TD v Play v Dasign v >> |  |  |
| 4        | Hedonometer<br>projects - About Instructions Words Blog Pres              | Hedorometer Hedoromete<br>s Papers Talks API Fund us logout             | © © ©                   | Introduction                           |  |
|          |   |   |                         | Measurement                            |  |
|          | The Wordshifte  | rator   |                         | Happiness                              |  |
|          |   |   |                         | Some motivation<br>Measuring emotional |  |
|          | Generate your own embeddable, shareable wordshifts for any pair of texts. |   |                         |  |  |
|          | Note: Texts must contain at least 1,000 words. Details about how this     | s works can be found here.  |                         | content<br>Hedonometer                 |  |
|          |   |   |                         | Analysis                               |  |
|          | Name of reference text:   | Name of comparison text:  |                         | Songs                                  |  |
|          | Reference Title   | Comparison Title  |                         | Blogs                                  |  |
|          | Reference text:   | Comparison text:  |                         | SOTU                                   |  |
|          | Paste Reference Text Here   | Paste Comparison Text Here  |                         | Geography<br>Movement                  |  |
|          |   |   |                         | Other Emotions                         |  |
|          |   |   |                         |  |  |
|          |   |   |                         | References                             |  |
|          |   |   |                         |  |  |
|          | Title:  |   |                         |  |  |
|          | Stopwords: Words to ignore, optional                                      |   |                         | :                                      |  |
|          | A comma separated list of words that do not<br>of these texts.            | make sense in the context   |                         | : 1 · 1 ·                              |  |
|          |   |   |                         |  |  |
|          | Language English ¢  |   |                         |  |  |
|          | Generate Wordshift  |   |                         |  |  |
| П.       | MITRE 🚳 🚾 🖓 🎞   |   |                         |  |  |
| Convis   | Cheveripy Viewel: 1978 1979 1979 1979 1979 1979 1979 1979                 |   |                         | 101                                    |  |
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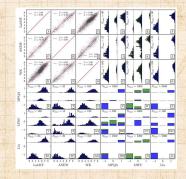
References



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### Quokka Labs

"Sentiment analysis methods for understanding large-scale texts: A case for using continuum-scored words and word shift graphs" Reagan, Tivnan, Williams, Danforth, and Dodds. EPJ Data Science, **6**, , 2017. <sup>[31]</sup>



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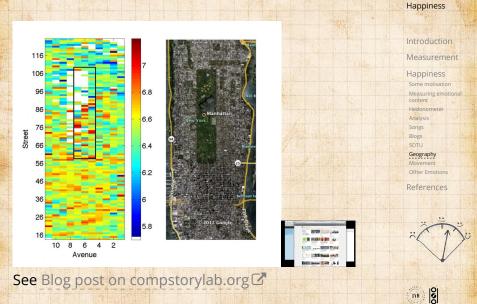
References

👶 Upshots: (1) do use wordshifts, and (2) do not use LIWC ...



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### Happiness in Manhattan:



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TELL US WHY

FOR A CHANCE TO WIN A TRIP FOR 2 TO JAMAICA!

INCLUDE #HAPPYNYC

Clear Channel Outdoor



TIMES SQUARE IS THE HAPPIEST PLACE IN NEW YORK!

AT FACEBOOK COMPCCOUTDOR

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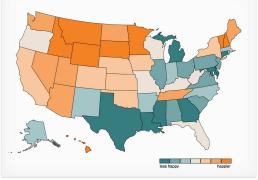
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# The Geography of Happiness:



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🚳 Online Appendices 🗹

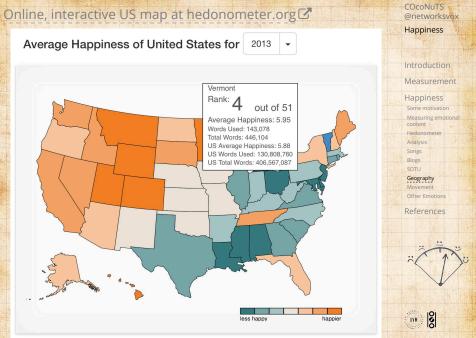
3

Much interesting and amusing press ...

line, interactive US map at hedonometer.org

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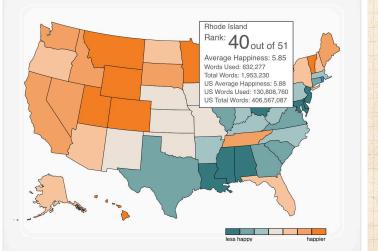


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Online, interactive US map at hedonometer.org

Average Happiness of United States for 20





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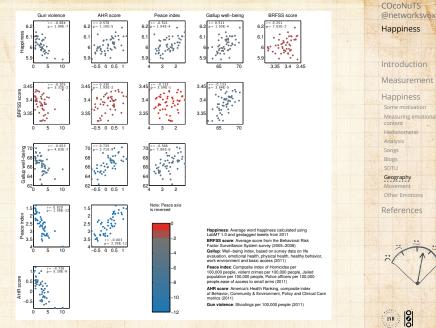
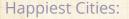
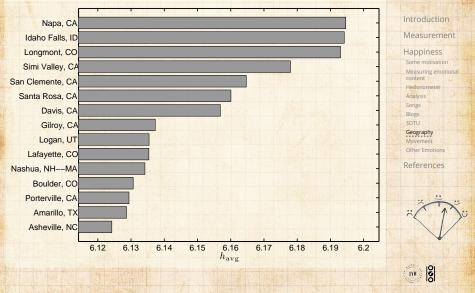


FIG. 2: Scatter plot matrix of correlations between different well-being measures. Points are colored by p-value, statistically insignificant correlations above p = 0.01 are shown in red. Spearman's r and p-value are reported in the inset.

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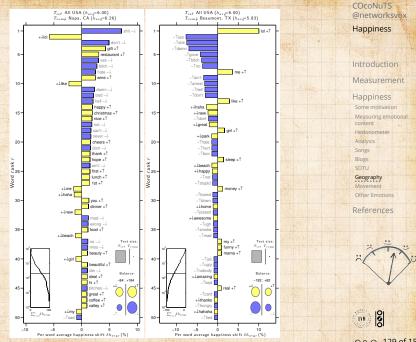
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### rednnneecckckkkkkkkk

2013/02/18 at 8:53 pm (Edit)



I've lived in quite a few places. The most recently Beaumont, TX. Its a pure hellhole. Hot, humid, trashy, terrible schools, corrupt government, lots of crime, no public parks or activities, terrible culture (other than crawfish boils), completely lacks diversity. This study confirms my suspicions that cities don't get any more miserable than this.

#### Reply

Blog post: Where is the happiest city in the US?





UVR SO

#### COcoNuTS @networksvox

#### Happiness



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UVN OO





"Happiness and the Patterns of Life: A Study of Geolocated Tweets" C Frank, Mitchell, Dodds, Danforth, and Dodds.

Nature Scientific Reports, 3, 2625, 2013. [14]

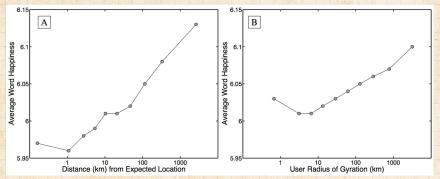


Figure 6 | (A) Average happiness of words written as a function of distance from an author's expected location, with tweets grouped into ten equally populated bins. Expressed happiness grows logarithmically with distance distance from expected location. (B) A similar trend is observed when individuals are grouped into ten equally populated bins according to their gyradius. Both trends persist through variations in binning and different measures of mobility.

### We grow fonder as we wander.

### Anger:

### Disgust:

word

tortured

murdered

holocaust

foundation

motherf\*\*ker

whore

asshole

killer

died

hawaii

arts

joy relaxing

relax

piano

presence

war

avg

4.16

3.74

3.67

3.66

3.56

3.55

3.54

3.48

3.40

1.06

1.06

1.04

1.04

1.04

1.04

1.04

1.00

σ 1.01

1.28

1.39

1.41

1.28

1.50

1.36

1.43

1.64

0.24

0.42

0.20

0.28

0.20

0.20

0.28

0.00

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| word            | avg  | σ    |  |
|-----------------|------|------|--|
| war             | 4.16 | 1.01 |  |
| torture         | 3.92 | 1.16 |  |
| murdered        | 3.84 | 1.14 |  |
| motherf☆☆ker    | 3.82 | 1.29 |  |
| anger           | 3.80 | 1.26 |  |
| killer          | 3.78 | 1.38 |  |
| fury            | 3.63 | 1.56 |  |
| bombing         | 3.58 | 1.39 |  |
|                 | :    | :    |  |
| play            | 1.06 | 0.31 |  |
| idea            | 1.06 | 0.31 |  |
| daughter-in-law | 1.06 | 0.24 |  |
| piano           | 1.06 | 0.31 |  |
| stars           | 1.06 | 0.24 |  |
| tasty           | 1.04 | 0.20 |  |
| thankful        | 1.02 | 0.14 |  |
| happy           | 1.00 | 0.00 |  |
|                 |      |      |  |

### Surprise:

motherf☆☆ker

murdered

bombing

death

lottery

torture

earthquake

fatal

slap

died

flag

b/c

pro beans

doors

stuart

johnson

word

avg

3.93

3.66

3.52

3.50

3.50

3.46

3.42

3.41

3.38

3.32

1.30

1.30

1.28

1.26

1.24

1.24

1.18

 $\sigma$ 

1.35

1.37

1.49

1.51

1.43

1.54

1.54

1.49

1.47

1.54

0.67

0.64

0.75

0.63

0.59

0.59

0.65

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#### Happiness

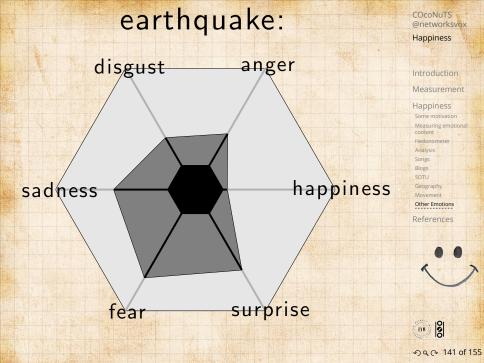
# Introduction Measurement Happiness Measuring emotional Hedonometer Analysis Blogs SOTU Geography Other Emotions References

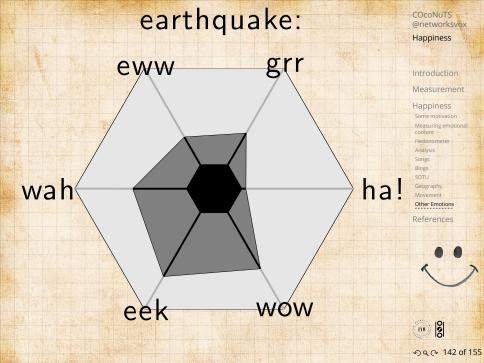
UVN OSO

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| word       | avg  | σ               |  |
|------------|------|-----------------|--|
| war        | 4.20 | 1.02            |  |
| tortured   | 4.18 | 1.39            |  |
| death      | 4.18 | 1.21            |  |
| killer     | 4.11 | 1.37            |  |
| murdered   | 4.06 | 1.10            |  |
| jail       | 3.90 | 1.08            |  |
| :          | i.   | ÷               |  |
| banana     | 1.08 | 0.34            |  |
| right      | 1.08 | 0.34            |  |
| properties | 1.08 | 0.27            |  |
| cute       | 1.06 | 0.24            |  |
| topic      | 1.06 | 0.24            |  |
| active     | 1.06 | 0.24            |  |
| wonderful  | 1.06 | 0.31            |  |
| dear       | 1.06 | 0.31            |  |
| bath       | 1.02 | 0.14            |  |
|            |      | manager and and |  |

Fear:

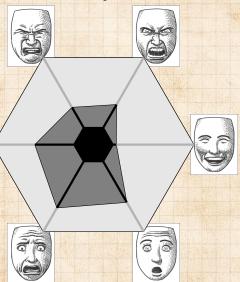




# earthquake:



Introduction



Images from Scott McCloud's "Making Comics." [27]

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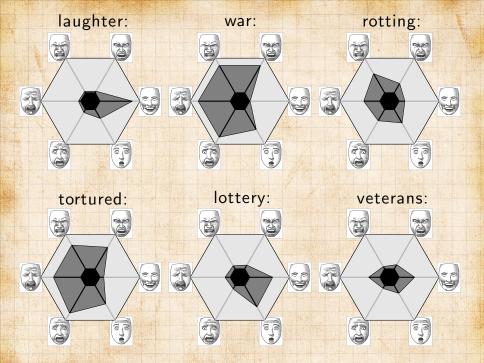
> Movement Other Emotions

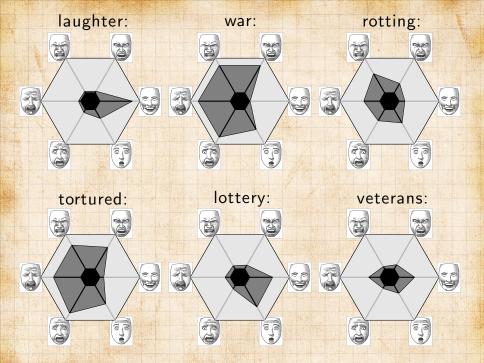
References



(in 18)

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### References I

[1] C. A. Bliss, I. M. Kloumann, K. D. Harris, C. M. Danforth, and P. S. Dodds. Twitter reciprocal reply networks exhibit assortativity with respect to happiness. Journal of Computational Science, 3:388–397, 2012. pdf 2

[2] M. M. Bradley and P. J. Lang. Affective norms for english words (anew): Stimuli, instruction manual and affective ratings. Technical report c-1, University of Florida, Gainesville, FL, 1999.

 [3] H. Chang.
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 Oxford University Press, 2004. COcoNuTS @networksvox Happiness

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### References II

E. M. Cody, A. J. Reagan, P. S. Dodds, and C. M. [4] Danforth. Public opinion polling with Twitter, 2016. Available online at https://arxiv.org/abs/1608.02024.pdf [5] T. Conner Christensen, L. Feldman Barrett, E. Bliss-Moreau, K. Lebo, and C. Kaschub. A practical guide to experience-sampling procedures. Journal of Happiness Studies, 4:53-78, 2003. M. Csikszentmihalvi. [6] Flow. Harper & Row, New York, 1990.

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- [7] M. Csikszentmihalyi, R. Larson, and S. Prescott. The ecology of adolescent activity and experience.
   Journal of Youth and Adolescence, 6:281–294, 1977.
- [8] E. Diener and M. Y. Chan. Happy people live longer: Subjective well-being contributes to health and longevity.
   <u>Applied Psychology: Health and Well-Being</u>, 3:1–43, 2011. pdf
- P. S. Dodds and C. M. Danforth.
   Measuring the happiness of large-scale written expression: Songs, blogs, and presidents.
   Journal of Happiness Studies, 11(4):441–456, 2009. pdf 2

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