Measuring the Happiness, Health, and Stories of Populations

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Complex Networks | @networksvox
CSYS/MATH 303, Spring, 2018

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Key papers:

"Measuring the happiness of large-scale written expression: Songs, blogs, and presidents.", Dodds and Danforth, Journal of Happiness Studies, **11**, 441–456, 2009. [10]

"Temporal patterns of happiness and information in a global social network: Hedonometrics and Twitter" Dodds et al., PLoS ONE, **6**, e26752, 2011. [12]

"Positivity of the English language", Kloumann et al., PLoS ONE, **7**, e29484, 2012. [24]

"Sentiment analysis methods for understanding large-scale texts: A case for using continuum-scored words and word shift graphs", Reagan et al., EPJ Data Science, **6**, , 2017. [32]

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

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."

Jon Stewart:

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

wildbluffmedia.com

- From the Daily Show⧉ (September 18, 2007; @5:13)
- The full inteview is here⧉.

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This is a Collateralized Debt Obligation:

Basic Science ≃ Describe + Explain:

Lord Kelvin (possibly):

- "To measure is to know."
- "If you cannot measure it, you cannot improve it."

But also:

- "There is nothing new to be discovered in physics now. All that remains is more and more precise measurement."
- "X-rays will prove to be a hoax."

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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 " a⧉ by Dava Sobel (2007). [34]
- Billionths of a second accuracy: Atomic clocks (Lord Kelvin, 1879)

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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." [22, 21]

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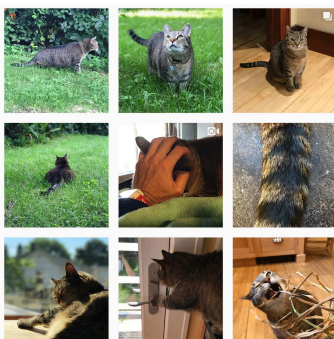Geography
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Measuring temperature was thought impossible:

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 rich manifestations of changes in temperature could be measured and quantified on a single numerical scale was scoffed at as impossible 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 [18] per @SilverVVulpes⧉: Also: Inventing Temperature, Hasok Chang, 2004 [4]
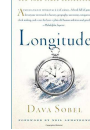
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Measuring temperature was thought impossible:

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 [18]
per @SilverVVulpes🔗: Also: Inventing Temperature, Hasok Chang, 2004 [4]
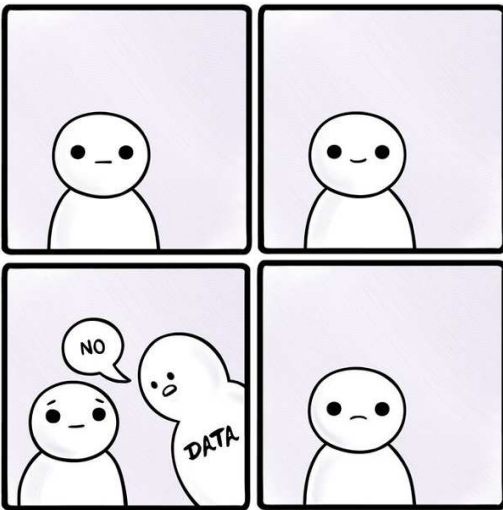
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What matters and what's measurable:

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Panometer—Three kinds of lexical meters:

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

Measuring Happiness:

Socrates et al.: eudaimonia [19]

Bentham: hedonistic calculus

Jefferson: ...the pursuit of happiness

Early drafts:

Happiness:

Even the odd modern economist is happy:

"Happiness" by Richard Layard [25]

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# What makes us happy?—Layard's summary:

Dominant factors:

- Family relationships
- Financial situation
- Work
- Community and Friends
- Health
- Personal Values
- Personal Freedom

Unimportant factors:

- Age
- Gender
- Education
- Inherent intelligence
- Looks

Desiring happiness—not just for boffins:

- Average people routinely report being happy is what they want most in life [25, 26, 9]
- And it matters: "Happy people live longer:…" Survey by Diener and Chan. [9]

A Plateau of Happiness

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.

HAPPINESS SCORES FROM 2000 POLL

Many countries, particularly those in Latin America, had higher marks for happiness than their economic situation would predict.

G.N.P. PER CAPITA (1995)

*Poll results for these countries were from 1995.
Source: Ronald Inglehart, "Human Beliefs and Values: A Cross-Cultural Sourcebook Based on the 1999-2002 Values Surveys"

National indices of well-being:

- Bhutan
- UK
- France
- Australia

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Some easy knocks:

"The Passionate State of Mind: And Other Aphorisms" a by Eric Hoffer (1954). [17]

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

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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:

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" a,↗
by Henepola Gunaratana (1992). [16]

"Flow" a,↗
by Mihaly Csikszentmihalyi (1990). [7]

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

Emotional content

So how does one measure
1. happiness?
2. levels of other emotional states?

Just ask people how happy they are.
- Experience sampling [6, 8, 7] (Csikszentmihalyi et al.)
- Day reconstruction [20] (Kahneman et al.)

But self-reporting has some drawbacks:
- relies on memory and self-perception
- induces misreporting [27]
- costly

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.

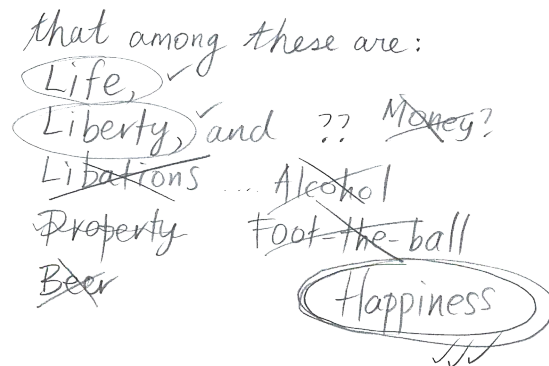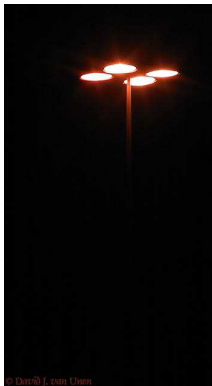Killingsworth and Gilbert, Science, 2010 [23]

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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:

- Transparent
- Fast
- Based on written expression
- Uses human evaluation

- Non-reactive
- Complementary to self-reported measures
- Improvable

We don't want to end up here:

Science

Policy

Or here: Facebook Lexicon Sentiment Analysis (2008)

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Facebook Lexicon Sentiment Analysis

Binary decision on emotional content

Sentiment score for "palin"
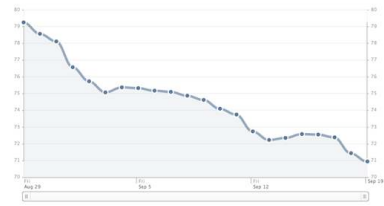
- Limitation: Sentiments are classified as either positive or negative.
- 'I like Sarah Palin' given same score as 'Sarah Palin's voice fills me with unbridled joy!'

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Measuring Emotional Content
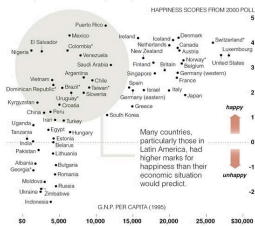
- Idea: Build on measures of the emotional content of individual words.
- Osgood et al. (1957) [31] identified a basis of three psychological variables as semantic differentials:
  - Valence: bad ↔ good
  - Arousal: passive ↔ active
  - Dominance: weak ↔ strong
  - Also often: Evaluation, Activity, and 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) [3]

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1999 ANEW study—three 1–9 scales: [3]

valence:

arousal:

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."

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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
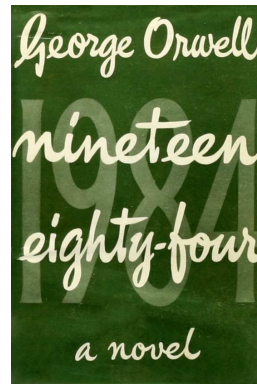- trauma/hostage/disgusted
- funeral/rape/suicide

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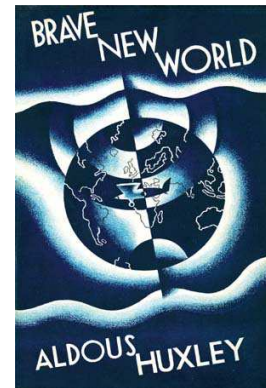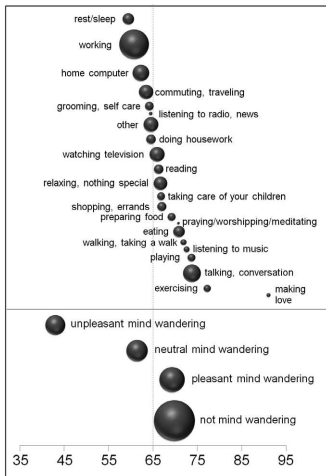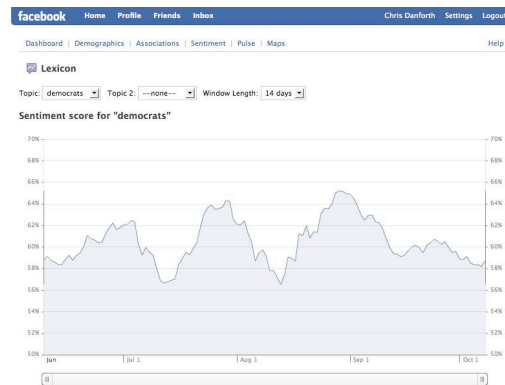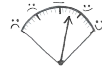Other Emotions
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Analysing text:

- Simplest measure for a text:

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

  where p_i is fractional abundance of word i and θ is average valence, arousal, or dominance for word i.

- Focus on happiness (valence), $\theta = h$.
- Average happiness typically falls between 5 and 7.

Song Lyrics—average happiness

Measuring the perceived happiness of a text:

Lyrics for Michael Jackson's Billie Jean

"She was more like a beauty queen from a movie scene.
⋮
And mother always told me, be careful who you love.
And be careful of what you do 'cause the lie becomes the truth.

Billie Jean is not my lover, She's just a girl who claims that I am the one.
⋮

ANEW words	v_k	f_k
k=1. love	8.72	1
2. mother	8.39	1
3. baby	8.22	3
4. beauty	7.82	1
5. truth	7.80	1
6. people	7.33	2
7. strong	7.11	1
8. young	6.89	2
9. girl	6.87	4
10. movie	6.86	1
11. perfume	6.76	1
12. queen	6.44	1
13. name	5.55	1
14. lie	2.79	1

$$v_{\text{text}} = \frac{\sum_k v_k f_k}{\sum_k f_k}$$

$v_{\text{Billie Jean}} = 7.1$

$v_{\text{Thriller}} = 6.3$

$v_{\text{Michael Jackson}} = 6.4$

- Temperature-like measure—large numbers only.
- Not meant to be used at level of sentence, paragraph, song, tweet, …
- Important: Social measure of sentiment.
- Later: see instrument is tunable.

Song Lyrics—measurement robustness

100 random subsets of 750 ANEW words

Daft Punk's "Around the World":

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.

Magic: Low entropy, high energy.

Song Lyrics—average happiness of genres:

- Gospel/Soul (6.91)
- Pop (6.69)
- Reggae (6.40)
- Rock (6.27)
- Rap/Hip−Hop (6.01)
- Punk (5.61)
- Metal/Industrial (5.10)

Happiness Word Shift Graph (early version):

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

Key:

Per word valence shift Δ_i

Word number i

♟ Word shifts are word clouds for grown ups.

Word data shift details:

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

♟ Measure difference in average happiness:
$h_{\text{avg}}^{(\text{comp})} - h_{\text{avg}}^{(\text{ref})}$

♟ Evident question: Which words contribute the most to this change?

♟ Break difference down by contributions from individual words:

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

♟ Must have: $\sum_i \delta h_{\text{avg},i} = \pm 100$

♟ Rank words by $|\delta h_{\text{avg},i}|$
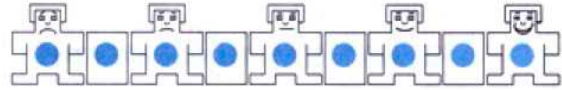
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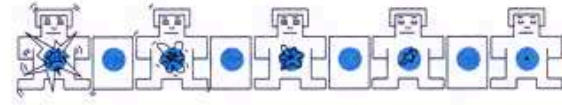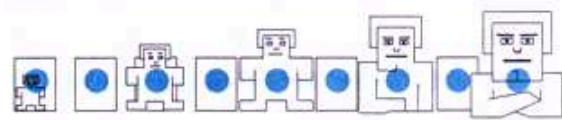Movement
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$+\uparrow$: Increased usage of relatively positive words—If a word is happier than text T_{ref} (+) and appears relatively more often in text T_{comp} (↑), then the contribution to the difference $h_{\text{avg}}^{(\text{comp})} - h_{\text{avg}}^{(\text{ref})}$ 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} (↓), then the contribution to the difference $h_{\text{avg}}^{(\text{comp})} - h_{\text{avg}}^{(\text{ref})}$ is also positive;

$+\downarrow$: 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 $h_{\text{avg}}^{(\text{comp})} - h_{\text{avg}}^{(\text{ref})}$ 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} (↑), then the contribution to the difference $h_{\text{avg}}^{(\text{comp})} - h_{\text{avg}}^{(\text{ref})}$ is also negative.

Word data shift details:

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

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

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

where

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

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

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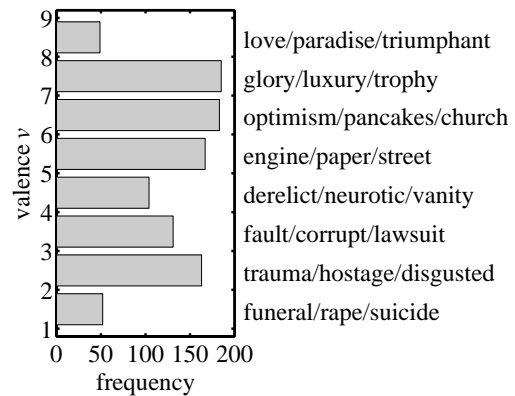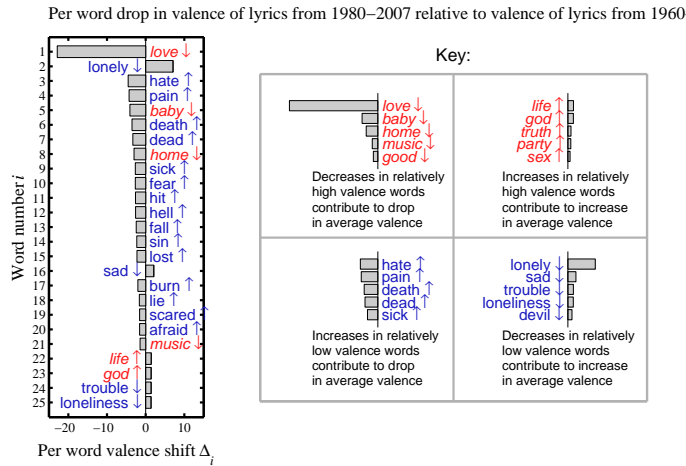Other Emotions

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

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

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

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

Rank	Artist	h_{avg}	Rank	Artist	h_{avg}
1	Slayer	4.80	26	Nine Inch Nails	5.34
2	Misfits	4.88	27	Sevendust	5.34
3	Staind	4.93	28	Annihilator	5.35
4	Slipknot	4.98	29	Biohazard	5.36
5	Darkthrone	4.98	30	Insane Clown Posse	5.36
6	Death	5.02	31	Megadeth	5.36
7	Black Label Society	5.05	32	Manowar	5.37
8	Pig	5.08	33	Zebrahead	5.38
9	Voivod	5.14	34	Danzig	5.39
10	Fear Factory	5.15	35	Acid Drinkers	5.40
11	Iced Earth	5.16	36	Dag Nasty	5.40
12	Simple Plan	5.16	37	Iron Maiden	5.40
13	Machine Head	5.17	38	Flotsam And Jetsam	5.41
14	Metallica	5.19	39	Powerman 5000	5.42
15	Dimmu Borgir	5.20	40	Anthrax	5.43
16	Mudvayne	5.21	41	Rhapsody	5.43
17	Linkin Park	5.22	42	Korn	5.43
18	Papa Roach	5.22	43	Rage	5.44
19	Audioslave	5.24	44	Accept	5.45
20	Rage Against The Machine	5.24	45	Esham	5.46
21	Cradle Of Filth	5.25	46	Blind Guardian	5.46
22	Dark Tranquility	5.26	47	White Zombie	5.47
23	Jack Off Jill	5.28	48	Helloween	5.50
24	Evanescence	5.30	49	W A S P	5.50
25	Twiztid	5.33	50	Green Day	5.50

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

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Text:	h_{avg}	Words with a similar score:
Soul/Gospel lyrics [11]	6.9	chocolate (6.88), leisurely (6.88), penthouse (6.81)
Pop lyrics [11]	6.7	dream (6.73), honey (6.73), sugar (6.74)
Dante's Paradise [?]	6.5	muffin (6.57), rabbit (6.57), smooth (6.58)
Tweets, 9/9/2008 to 12/31/2010	6.4	thought (6.39), face (6.39), blond (6.42)
Rock lyrics [11]	6.3	church (6.28), tree (6.32), air (6.34)
Enron Emails [?]	6.2	clouds (6.18), alert (6.20), computer (6.24)
State of the Union Messages [11]	6.1	grass (6.12), idol (6.12), bottle (6.15)
New York Times (1987–2007) [33]	6.0	hotel (6.00), tennis (6.02), wonder (6.03)
Blogs [11]	5.8	owl (5.80), whistle (5.81), humble (5.86)
Dante's Inferno [?]	5.5	glacier (5.50), repentant (5.53), mischief (5.57)
Heavy Metal lyrics [11]	5.4	lamp (5.41), elevator (5.44), truck (5.47)

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.

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Lexicon Valley, Episode #62, June 17, 2015

Mike Vuolo and Bob Garfield.

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

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wefeelfine.org:

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Data sets:

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

Created by Jonathan Harris & Sep Kamvar

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wefeelfine.org:
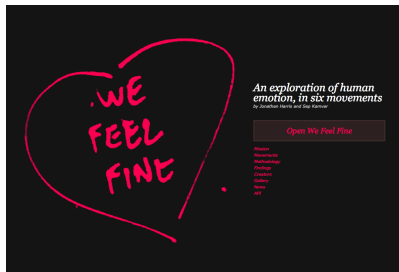
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Blogs—Overall trend

average happiness h_{avg}

US Election 11/4
US Inauguration 1/20
9/10
9/11
9/10
Michael Jackson

A S O N D J F M A M J J A S O N D J F M A M J J A S O N D J F M A M J J A S O N D J F M A M J J A S O N D J F M A

2005 2006 2007 2008 2009 2010

T_{ref}: born in 1960-1969 (h_{avg}=5.96)
T_{comp}: 14 years old (h_{avg}=5.55)

Word rank r

$-\uparrow$sick
$-\uparrow$hate
$-\uparrow$stupid
$-\uparrow$sad
happy $+\uparrow$
love $+\uparrow$
$-\uparrow$depressed
$-\uparrow$bored
$-\uparrow$lonely
$-\uparrow$alone
$-\uparrow$mad
$-\uparrow$pain
$+\downarrow$life
loved $+\uparrow$
$-\uparrow$upset
$-\uparrow$fat
fun $+\uparrow$
$-\uparrow$dead
$-\uparrow$scared
$-\uparrow$terrible
friend $+\uparrow$
people $+\uparrow$
$-\uparrow$confused
time $-\downarrow$
$-\uparrow$hurt

Text size:
T_{ref} T_{comp}

Balance:
$-169 : +69$
$+\uparrow$
$-\downarrow$

$\sum_{i=1}^{r} \delta h_{avg\,i}$

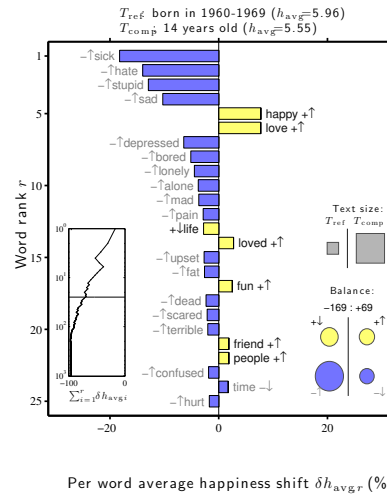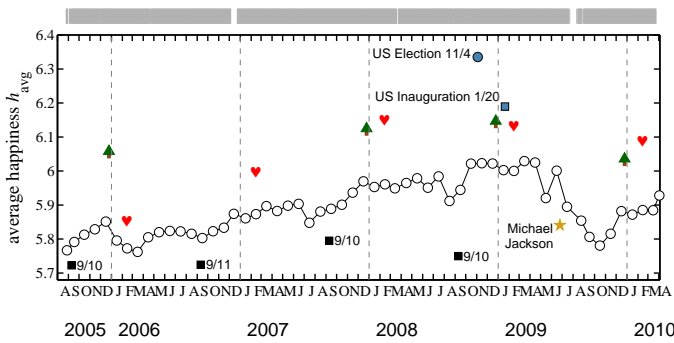Per word average happiness shift $\delta h_{avg\,r}$ (%)

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From wefeelfine.org by Jonathan Harris & Sep Kamvar

valence (v)

blogger age

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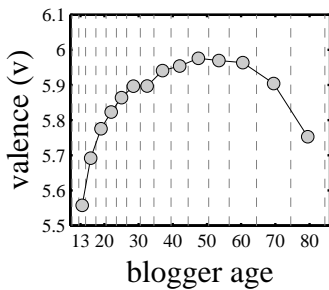
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Blogs—Latitude

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.'
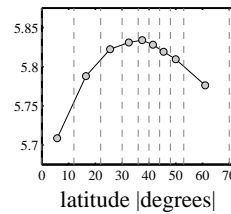

latitude |degrees|

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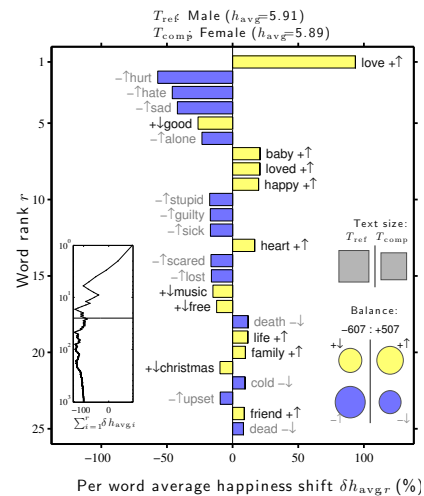Panometrics
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Blogs—Age

🐾 Self-report studies find little variation in happiness with age [13, 14]

🐾 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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T_{ref}: Male (h_{avg}=5.91)
T_{comp}: Female (h_{avg}=5.89)

Word rank r

love $+\uparrow$
$-\uparrow$hurt
$-\uparrow$hate
$-\uparrow$sad
$+\downarrow$good
$-\uparrow$alone
baby $+\uparrow$
loved $+\uparrow$
happy $+\uparrow$
$-\uparrow$stupid
$-\uparrow$guilty
$-\uparrow$sick
heart $+\uparrow$
$-\uparrow$scared
$-\uparrow$lost
$+\downarrow$music
$+\downarrow$free
death $-\downarrow$
life $+\uparrow$
family $+\uparrow$
$+\downarrow$christmas
cold $-\downarrow$
$-\uparrow$upset
friend $+\uparrow$
dead $-\downarrow$

Text size:
T_{ref} T_{comp}

Balance:
$-607 : +507$
$+\uparrow$
$-\downarrow$

$\sum_{i=1}^{r} \delta h_{avg\,i}$

Per word average happiness shift $\delta h_{avg\,r}$ (%)
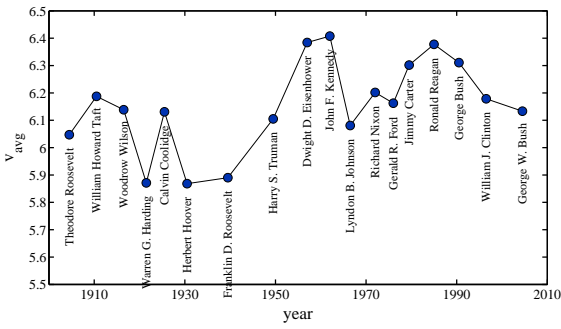
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labMT 1.0:
language assessment by Mechanical Turk

Four corpora:

- Twitter
- Google Books
- Music Lyrics
- New York Times

5000 most frequently used words for each corpus.

10,222 words, 50 evaluations each, 1–9 scale: [28]
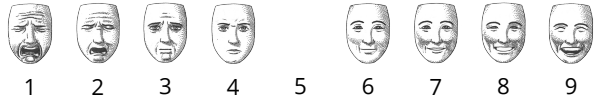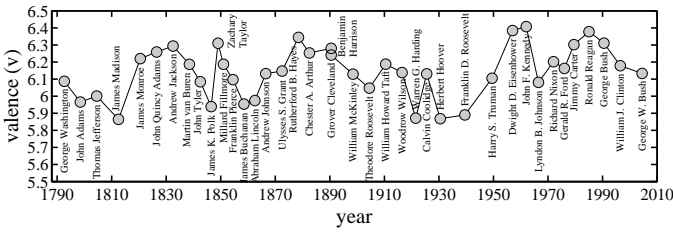
| 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |

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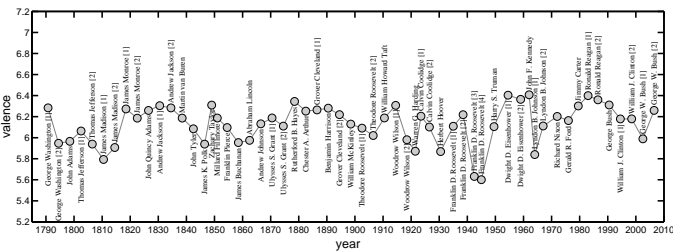Other Emotions
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valence rank	word	valence	std dev	twitter rank	g-books rank	nyt rank	lyrics rank
1	laughter	8.50	0.93	3600	–	–	1728
2	happiness	8.44	0.97	1853	2458	–	1230
3	love	8.42	1.11	25	317	328	23
4	happy	8.30	0.99	65	1372	1313	375
5	laughed	8.26	1.16	3334	3542	–	2332
6	laugh	8.22	1.37	1002	3998	4488	647
7	laughing	8.20	1.11	1579	–	–	1122
8	excellent	8.18	1.10	1496	1756	3155	–
9	laughs	8.18	1.16	3554	–	–	2856
10	joy	8.16	1.06	988	2336	2723	809
11	successful	8.16	1.08	2176	1198	1565	–
12	win	8.12	1.08	154	3031	776	694
13	rainbow	8.10	0.99	2726	–	–	1723
14	smile	8.10	1.02	925	2666	2898	349
15	won	8.10	1.22	810	1167	439	1493
16	pleasure	8.08	0.97	1497	1526	4253	1398
17	smiled	8.08	1.07	–	3537	–	2248
18	rainbows	8.06	1.36	–	–	–	4216
19	winning	8.04	1.05	1876	–	1426	3646
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	–	–	–
25	congratulations	8.00	1.63	2246	–	–	–
26	weekend	8.00	1.29	527	317	833	2256
27	celebrate	7.98	1.15	1606	–	3574	2108
28	comedy	7.98	1.15	1444	2762	2566	–
29	jokes	7.98	0.98	2812	–	–	3808
30	rich	7.98	1.32	1625	1221	1469	890

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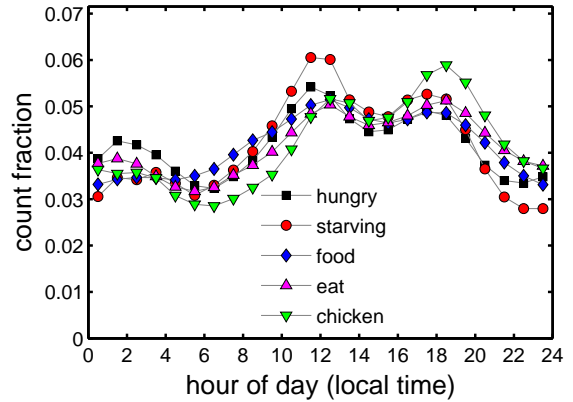Other Emotions
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valence rank	word	valence	std dev	twitter rank	g-books rank	nyt rank	lyrics rank
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
10193	violence	1.86	1.05	4299	1724	1238	2016
10194	cruel	1.84	1.15	2963	–	–	1447
10195	cry	1.84	1.28	1028	3075	–	226
10196	failed	1.84	1.00	2645	1618	1276	2920
10197	sickness	1.84	1.18	4735	–	–	3782
10198	abused	1.83	1.31	–	–	–	4589
10199	tortured	1.82	1.42	–	–	–	4693
10200	fatal	1.80	1.53	–	4089	–	3724
10201	killings	1.80	1.54	–	–	4914	–
10202	murdered	1.80	1.63	–	–	–	4796
10203	war	1.80	1.41	468	175	291	462
10204	kills	1.78	1.23	2459	–	–	2857
10205	jail	1.76	1.02	1642	–	2573	1619
10206	terror	1.76	1.00	4625	4117	4048	2370
10207	die	1.74	1.19	418	730	2605	143
10208	killing	1.70	1.36	1507	4428	1672	998
10209	arrested	1.64	1.01	2435	4474	1435	–
10210	deaths	1.64	1.14	–	–	2974	–
10211	raped	1.64	1.43	–	–	–	4528
10212	torture	1.58	1.05	3175	–	–	3126
10213	died	1.56	1.20	1223	866	208	826
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	–	–	3192	–
10220	rape	1.44	0.79	3133	–	4115	2977
10221	suicide	1.30	0.84	2124	4707	3319	2107
10222	terrorist	1.30	0.91	3576	–	3026	–

std dev rank	word	valence	std dev	twitter rank	g-books rank	nyt rank	lyrics rank
1	f♩♩king	4.64	2.93	448	–	–	620
2	f☆☆kin	3.86	2.74	1077	–	–	688
3	f☆☆ked	3.56	2.71	1840	–	–	904
4	pussy	4.80	2.66	2019	–	–	949
5	whiskey	5.72	2.64	–	–	–	2208
6	slut	3.57	2.63	–	–	–	4071
7	cigarettes	3.31	2.60	–	–	–	3279
8	f☆☆k	4.14	2.58	322	–	–	185
9	mortality	4.38	2.55	–	3960	–	–
10	cigarette	3.09	2.52	–	–	–	2678
11	motherf☆☆kers	2.51	2.47	–	–	–	1466
12	churches	5.70	2.46	–	2281	–	–
13	motherf☆☆king	2.64	2.46	–	–	–	2910
14	capitalism	5.16	2.45	–	4648	–	–
15	porn	4.18	2.43	1801	–	–	–
16	summer	6.40	2.39	896	1226	721	590
17	beer	5.92	2.39	839	4924	3960	1413
18	execution	3.10	2.39	–	2975	–	–
19	wines	6.28	2.37	–	–	3316	–
20	zombies	4.00	2.37	4708	–	–	–
21	aids	4.28	2.35	2983	3996	1197	–
22	capitalist	4.84	2.34	–	4694	–	–
23	revenge	3.71	2.34	–	–	–	2766
24	mcdonalds	5.98	2.33	3831	–	–	–
25	beatles	6.44	2.33	3797	–	–	–
26	islam	4.68	2.33	–	4514	–	–
27	pay	5.30	2.32	627	769	460	499
28	alcohol	5.20	2.32	2787	2617	3752	3600
29	muthaf☆☆kin	3.00	2.31	–	–	–	4107
30	christ	6.16	2.31	2509	909	4238	1526
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮

Twitter—living in the now:

Quantifying the quotidian.

Twitter—living in the now:

Makes the unexpected believable...

The happiest distribution:
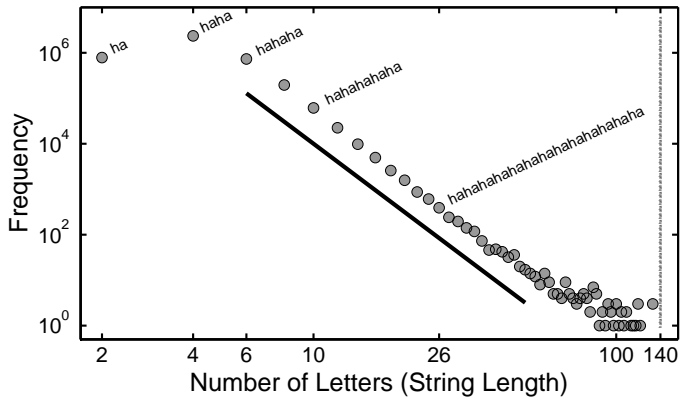

Work by Tyler Gray et al., unpublished.

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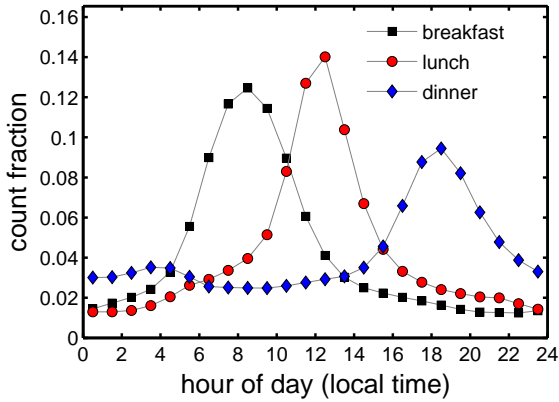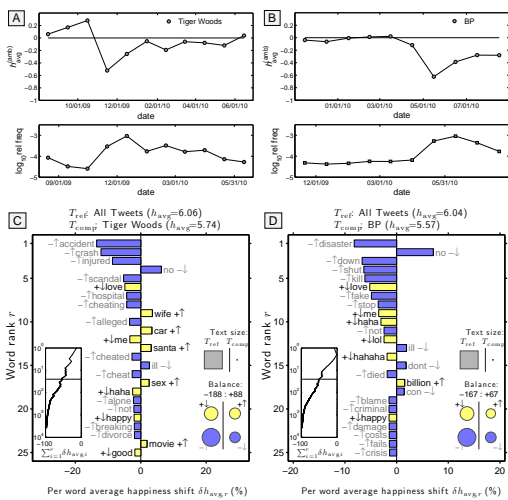Movement
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Tom Sherwood @tomsherwood   1 Oct

Mississippi congressman moves gate and lets 92 WW2 vets see their monument. pic.twitter.com/541GvLLITz

Retweeted 442 times

Expand

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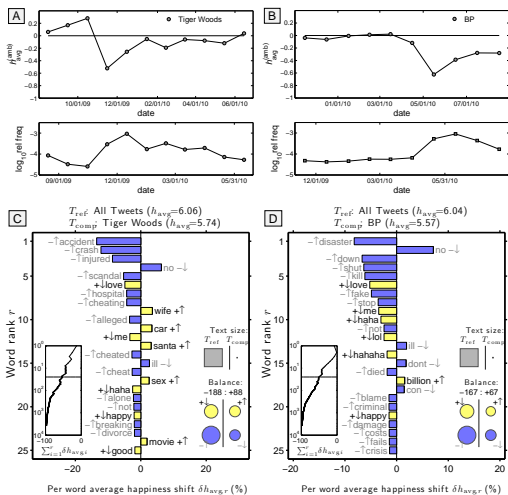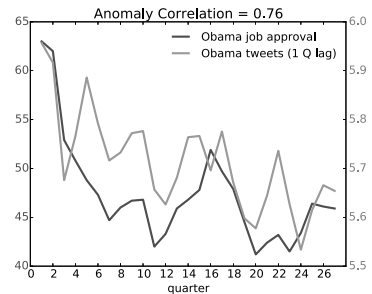SOTU
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{"created_at":"Tue Oct 01 15:59:45 +0000 2013","id":"385071546731347968","id_str":"385071546731347968","text":"RT @tomsherwood: Mississippi congressman moves gate and lets 92 WW2 vets see their monument. http:\/\/t.co\/54lGvLLTz","source":"\u003cTweetDeck\u003e"...

Individual tweets have all kinds of potential impact:

Meme come true: Fleetwood Mac re-enter US

Mentions of CIA on Twitter, end of 2011:

See story here ⧉ for example [slate].

The daily unravelling of the human mind:

"Social Scientists wade into the Tweet stream" by Greg Miller,
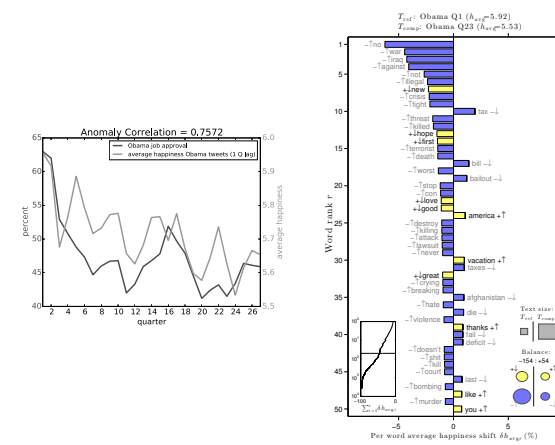Science, **333**, 1814–1815, 2011 [29]

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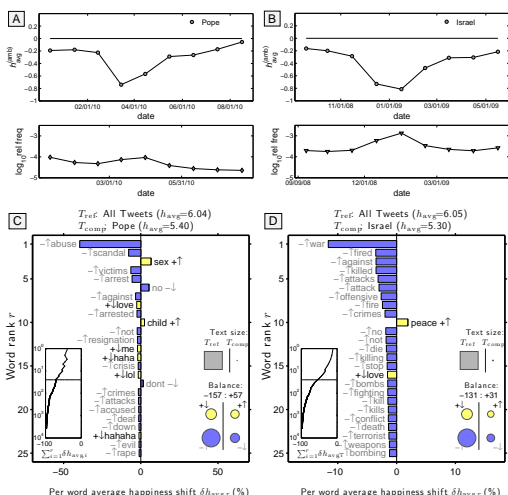Panometrics

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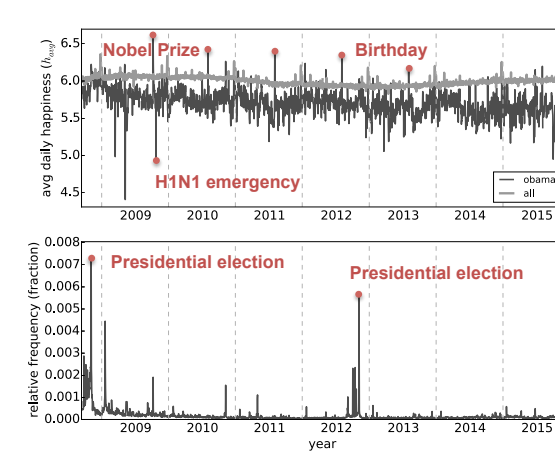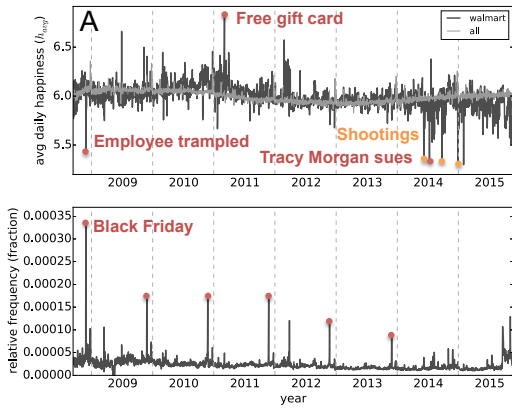Other Emotions

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Text element and context correlate in happiness scores:

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

Simpson lexical size, N_S:

Word	$h_{avg}^{(amb)}$	Total Tweets	$h_{avg}^{(norm)}$		Word	$h_{avg}^{(amb)}$	Total Tweets	$h_{avg}^{(norm)}$
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				52. Jon Stewart			
3. vegan	+0.315				53. school			
4. :)	+0.274				54. Lehman Brothers			
5. family	+0.251				55. them			
6. :-)					56. right			
7. our	+0.207				57. woman			
8. win					58. left			
9. vacation	+0.200				59. me			
10. party	+0.170				60. election			
11. love	+0.164				61. Sarah Palin			
12. friends					62. no			
13. hope	+0.149				63. rain			
14. coffee					64. climate			
15. cash	+0.146				65. gay			
16. sun	+0.144				66. lose			
17. income	+0.137				67. they			
18. summer					68. oil			
19. church	+0.131				69. cold			
20. Valentine					70. I feel			
21. Stephen Colbert	+0.126				71. man			
22. USA					72. Republican			
23. !	+0.106				73. sad			
24. winter					74. gas			
25. God	+0.099				75. economy			
26. hot					76. Obama			
27. :)	+0.094				77. Democrat			
28. Jesus	+0.094				78. Congress			
29. today	+0.092				79. hell			
30. kiss					80. sick			
31. yes	+0.056				81. Muslim			
32. tomorrow					82. war			
33. you	+0.052				83. Pope			
34. heaven					84. hate			
35. ;-)	+0.041				85. Glenn Beck			
36. we					86. Islam			
37. yesterday	+0.033				87. George Bush			
38. dark	+0.031				88. Goldman Sachs			
39. ?	-0.030				89. depressed			
40. RT	+0.018				90. Senate			
41. Michael Jackson	+0.018				91. BP	-0.355		
42. night	+0.014				92. gun			
43. life	+0.012				93. drugs	-0.382		
44. health	-0.000				94. headache			
45. sex	-0.008				95. :(	-0.455		
46. work	-0.010				96. :-(			
47. girl	-0.010				97. Afghanistan	-0.703		
48. boy	-0.026				98. mosque			
49. I	-0.048				99. flu	-0.735		
50. commute	-0.048				100. Iraq	-0.773		

Bailout of the U.S. financial system:
T_{ref}: 7 days before and after (h_{avg}=6.00).
T_{comp}: Monday, 2008/09/29 (h_{avg}=5.95).

Royal Wedding of Prince William & Catherine Middleton:
T_{ref}: 7 days before and after (h_{avg}=5.98).
T_{comp}: Friday, 2011/04/29 (h_{avg}=6.04).

Death of Osama Bin Laden:
T_{ref}: 7 days before and after (h_{avg}=5.98).
T_{comp}: Monday, 2011/05/02 (h_{avg}=5.89).

Ambient happiness:
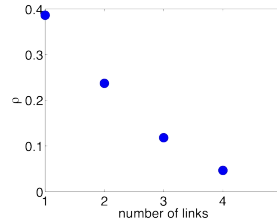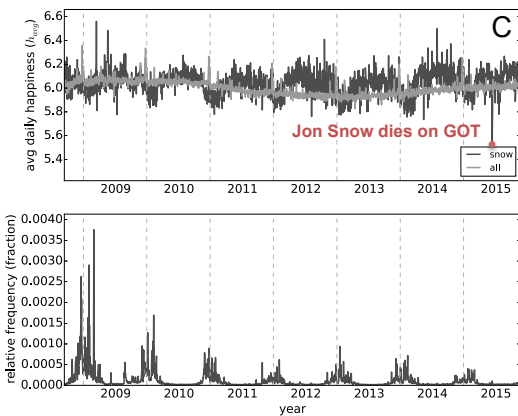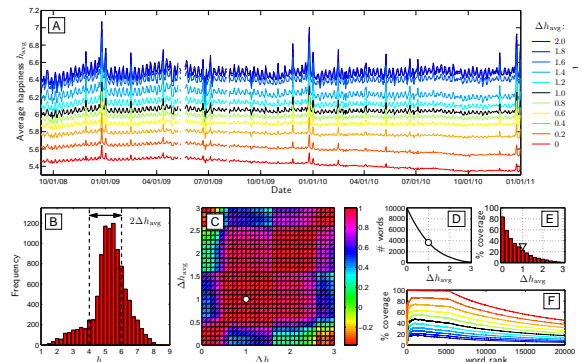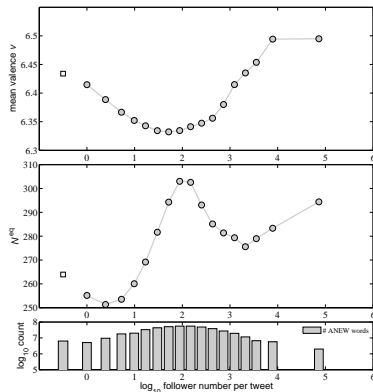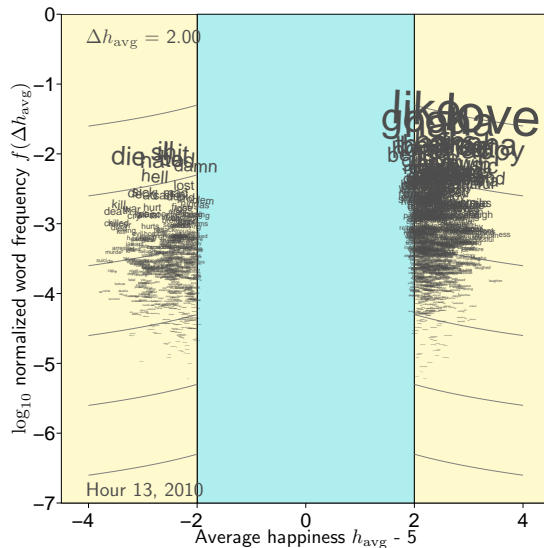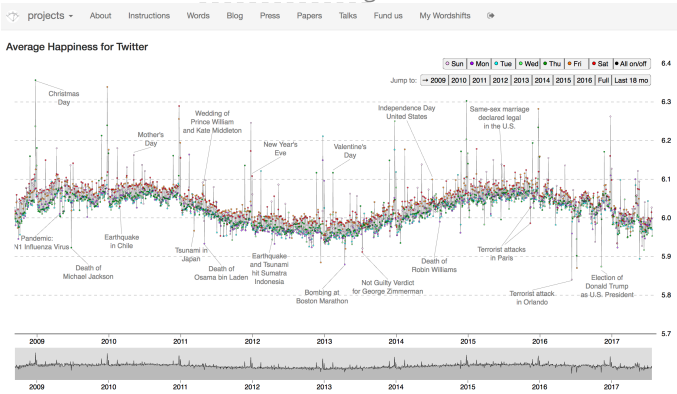

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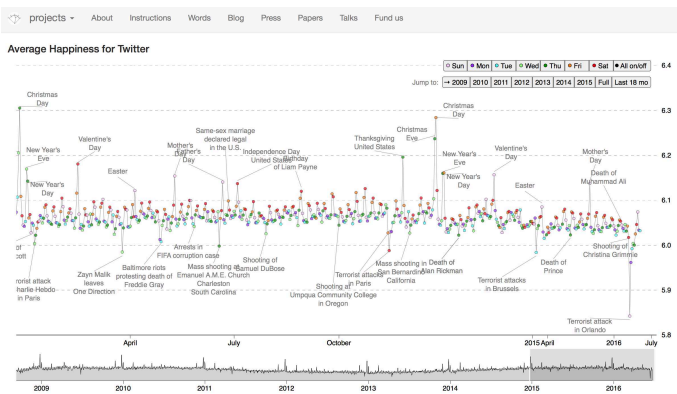Movement
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Ambient happiness:

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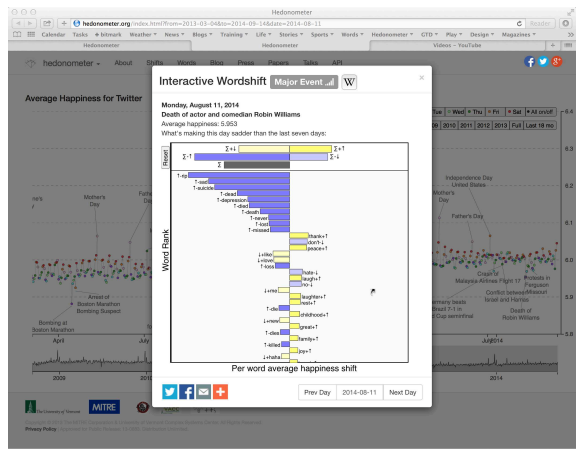Movement
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Slide A/B (Tiger Woods / BP): $\eta_{avg}^{(amb)}$ vs date; $\log_{10}$ rel freq vs date

Slide C: T_{ref}: All Tweets (h_{avg}=6.06), T_{comp}: Tiger Woods (h_{avg}=5.74)
Slide D: T_{ref}: All Tweets (h_{avg}=6.04), T_{comp}: BP (h_{avg}=5.57)

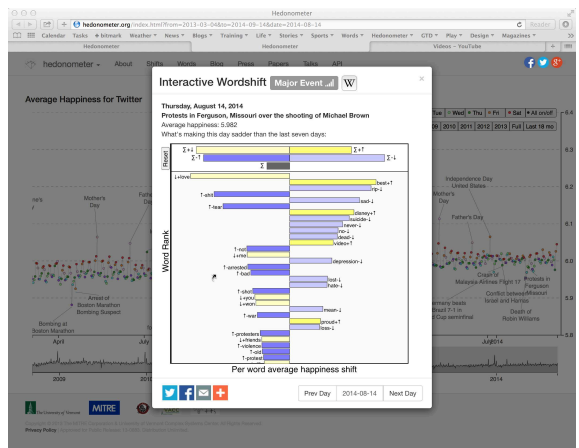Per word average happiness shift $\delta h_{avg,r}$ (%)

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"Public opinion polling with Twitter"
Cody et al.,
Available online at
https://arxiv.org/abs/1608.02024, 2016. [5]

Anomaly Correlation = 0.76
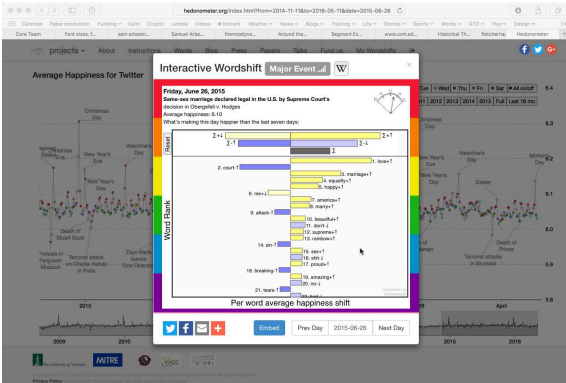
Obama job approval
Obama tweets (1 Q lag)

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Slide C: T_{ref}: All Tweets (h_{avg}=6.06), T_{comp}: Tiger Woods (h_{avg}=5.74)
Slide D: T_{ref}: All Tweets (h_{avg}=6.04), T_{comp}: BP (h_{avg}=5.57)

Per word average happiness shift $\delta h_{avg,r}$ (%)
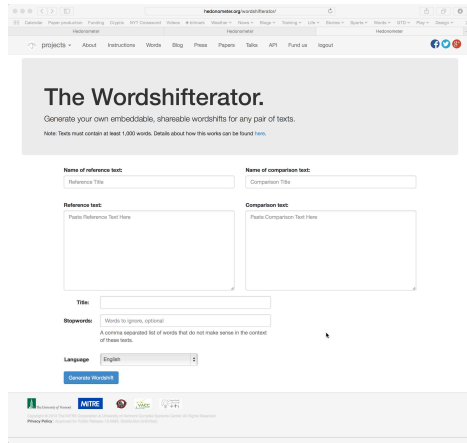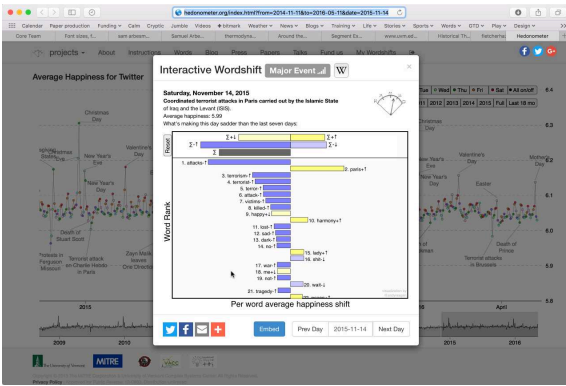

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Twitter's Feels predict Obama's Approval Rating:

Anomaly Correlation = 0.7572

Obama job approval
average happiness Obama tweets (1 Q lag)

T_{ref}: Obama Q1 (h_{avg}=5.92)
T_{comp}: Obama Q23 (h_{avg}=5.53)

Per word average happiness shift $\delta h_{avg,r}$ (%)

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Slide A/B (Pope / Israel)

Slide C: T_{ref}: All Tweets (h_{avg}=6.04), T_{comp}: Pope (h_{avg}=5.40)
Slide D: T_{ref}: All Tweets (h_{avg}=6.05), T_{comp}: Israel (h_{avg}=5.30)

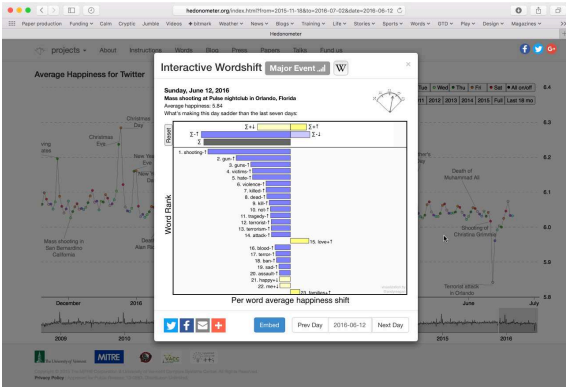Per word average happiness shift $\delta h_{avg,r}$ (%)
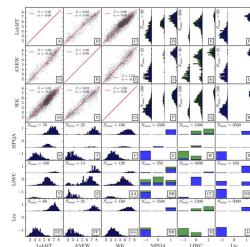

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Ambient happiness for "Obama":

avg daily happiness (h_{avg})

Nobel Prize   Birthday
H1N1 emergency
obama
all

relative frequency (fraction)

Presidential election   Presidential election

Ambient happiness for "Walmart":

Ambient happiness for "snow":

Dunbar number action:

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

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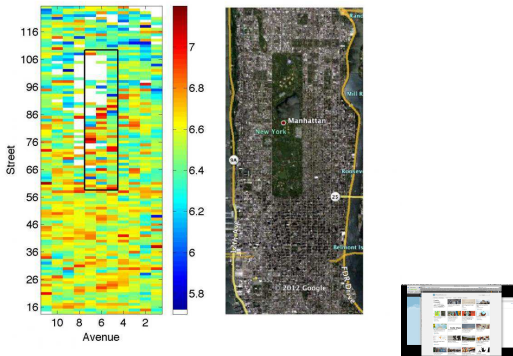References

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"Twitter reciprocal reply networks exhibit assortativity with respect to happiness"⧉
Bliss, Kloumann, Harris, Danforth, and Dodds.
Journal of Computational Science, **3**, 388–397, 2012. [2]

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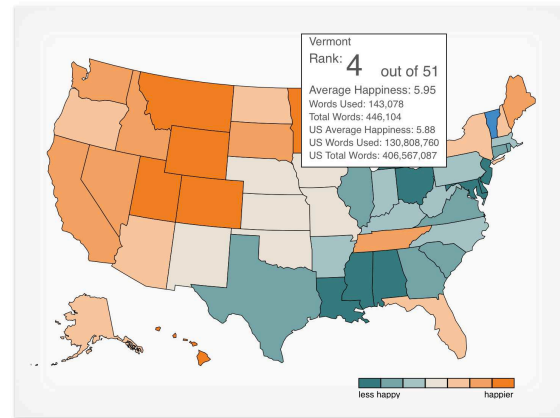- Not a test of contagion ...

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The very surprising tunable hedonometer:

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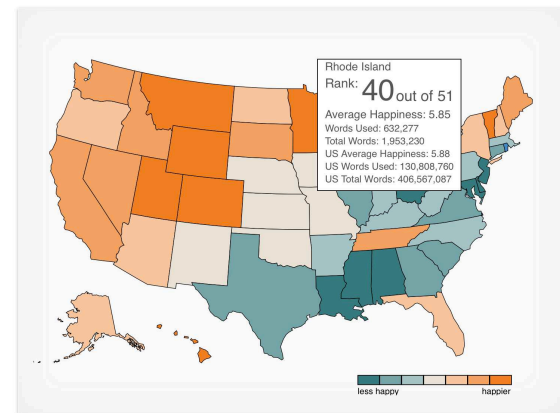Other Emotions

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Online instrument: hedonometer.org ↗

🖐 Machine: @andyreagan ↗

🖐 Planned happiness versus tragedies.

hedonometer.org ↗ —word shifts:
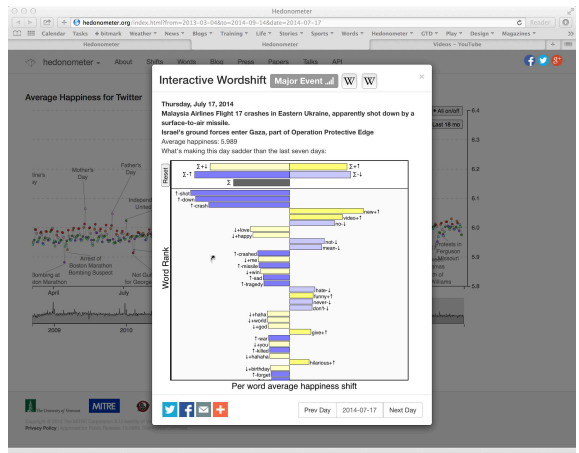
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🖐 Machine: @andyreagan ↗

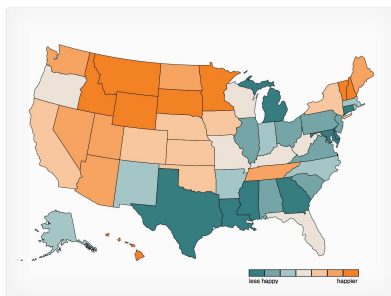## hedonometer.org ↗ —word shifts:

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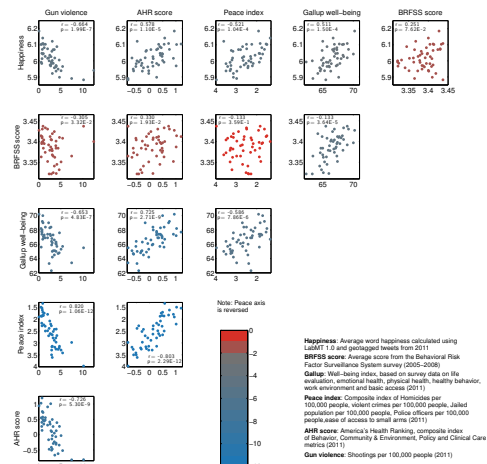Geography
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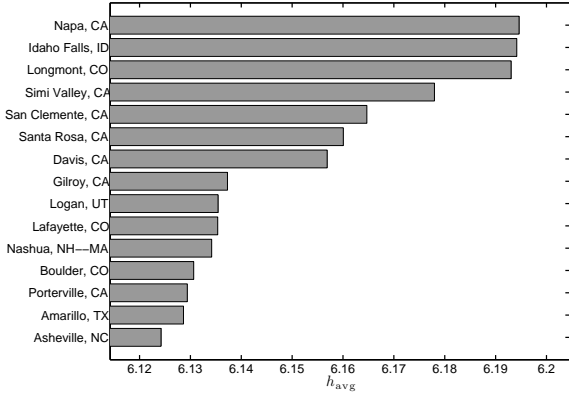Geography
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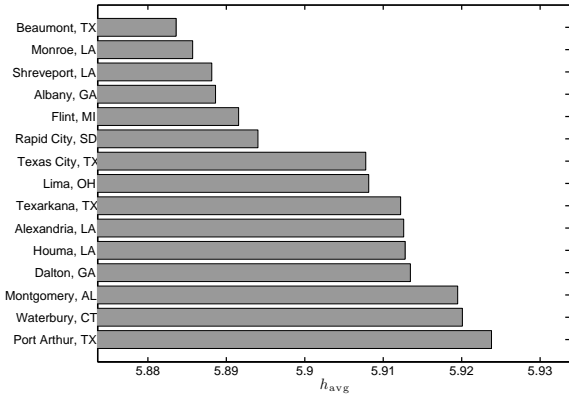
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Make your own: ⬈

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"Sentiment analysis methods for understanding large-scale texts: A case for using continuum-scored words and word shift graphs" ⬈
Reagan, Tivnan, Tivnan, Danforth, and Danforth.
EPJ Data Science, **6**, , 2017. [32]

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

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Happiness in Manhattan:

See Blog post on compstorylab.org ↗

The Geography of Happiness:

- Mitchell et al., PLoS ONE, 2013. [30]
- It's a paper that tweets: @geographyofhapp ↗
- Online **App**endices ↗
- Much interesting and amusing press ...
- Online, interactive US map at hedonometer.org ↗

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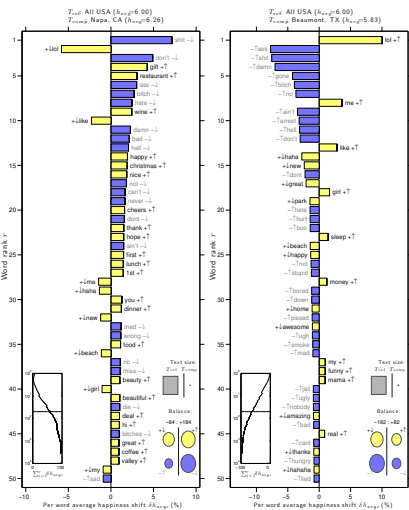 Other Emotions
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Online, interactive US map at hedonometer.org ↗

Average Happiness of United States for 2013 ▾

Vermont
Rank: **4** out of 51
Average Happiness: 5.95
Words Used: 143,078
Total Words: 446,104
US Average Happiness: 5.88
US Words Used: 130,808,760
US Total Words: 406,567,087
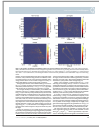
less happy   happier

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Online, interactive US map at hedonometer.org ↗

Average Happiness of United States for 2013 ▾

Rhode Island
Rank: **40** out of 51
Average Happiness: 5.85
Words Used: 632,277
Total Words: 1,953,230
US Average Happiness: 5.88
US Words Used: 130,808,760
US Total Words: 406,567,087

less happy   happier

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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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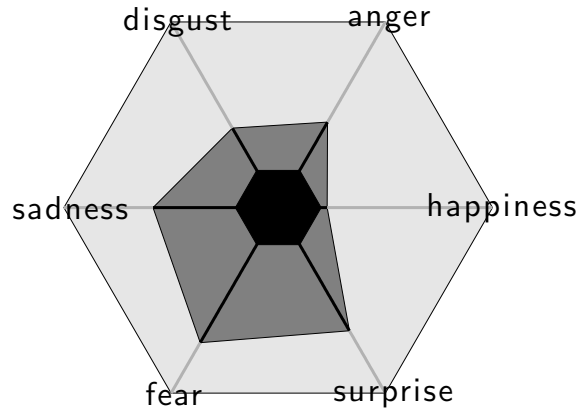 Geography
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Happiest Cities:

Saddest Cities (Sorry Beaumont):

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?

The Bayou
Biting satire, commentary & opinion Southeast-Texas style

Eggheads find Beaumont IS the 'saddest city'
February 19, 2013 at 10:41 am by gstor

That's what some eggheads at the Department of Mathematics and Statistics at the University of Vermont have determined.

The method to their madness? Twitter. They say they plotted 10 million geotagged Tweets in 2011.

"Happiness and the Patterns of Life: A Study of Geolocated Tweets"

Frank, Mitchell, Dodds, Danforth, and Danforth.

Nature Scientific Reports, **3**, 2625, 2013. [15]
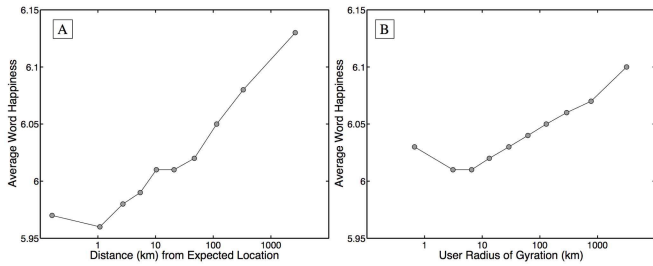
Figure 6 | (A) Average happiness of words written as a function of distance from an author's expected location, with tweets grouped into ten equally populated bins. Expressed happiness grows logarithmically with distance distance from expected location. (B) A similar trend is observed when individuals 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:

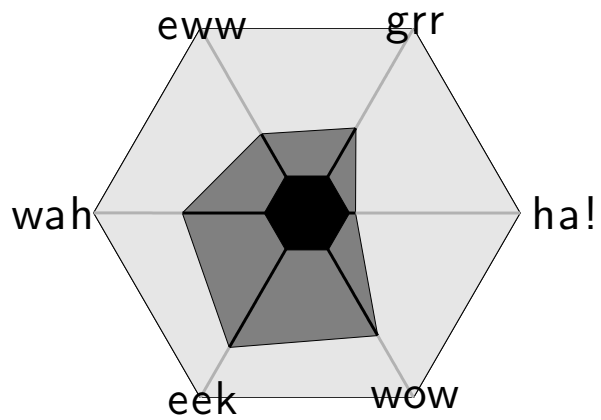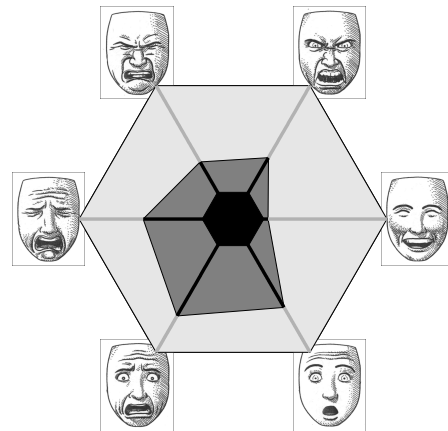| word	avg	σ
war	4.16	1.01
torture	3.92	1.16
murdered	3.84	1.14
motherf☆☆ker	3.82	1.29
anger	3.80	1.26
killer	3.78	1.38
fury	3.63	1.56
bombing	3.58	1.39
⋮	⋮	⋮
play	1.06	0.31
idea	1.06	0.31
daughter-in-law	1.06	0.24
piano	1.06	0.31
stars	1.06	0.24
tasty	1.04	0.20
thankful	1.02	0.14
happy	1.00	0.00

Disgust:

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word	avg	σ
war	4.16	1.01
tortured	3.74	1.28
whore	3.67	1.39
murdered	3.66	1.41
asshole	3.56	1.28
killer	3.55	1.50
motherf☆☆ker	3.54	1.36
died	3.48	1.43
holocaust	3.40	1.64
⋮	⋮	⋮
hawaii	1.06	0.24
arts	1.06	0.42
joy	1.04	0.20
relaxing	1.04	0.28
foundation	1.04	0.20
relax	1.04	0.20
piano	1.04	0.28
presence	1.00	0.00

Fear:

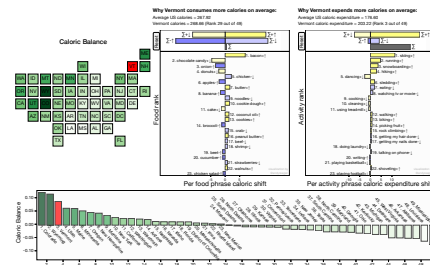| word	avg	σ
war	4.20	1.02
tortured	4.18	1.39
death	4.18	1.21
killer	4.11	1.37
murdered	4.06	1.10
jail	3.90	1.08
⋮	⋮	⋮
banana	1.08	0.34
right	1.08	0.34
properties	1.08	0.27
cute	1.06	0.24
topic	1.06	0.24
active	1.06	0.24
wonderful	1.06	0.31
dear	1.06	0.31
bath	1.02	0.14

Surprise:

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word	avg	σ
motherf☆☆ker	3.93	1.35
murdered	3.66	1.37
bombing	3.52	1.49
death	3.50	1.51
fatal	3.50	1.43
lottery	3.46	1.54
torture	3.42	1.54
slap	3.41	1.49
died	3.38	1.47
earthquake	3.32	1.54
⋮	⋮	⋮
flag	1.30	0.67
doors	1.30	0.64
b/c	1.28	0.75
stuart	1.26	0.63
pro	1.24	0.59
beans	1.24	0.59
johnson	1.18	0.65

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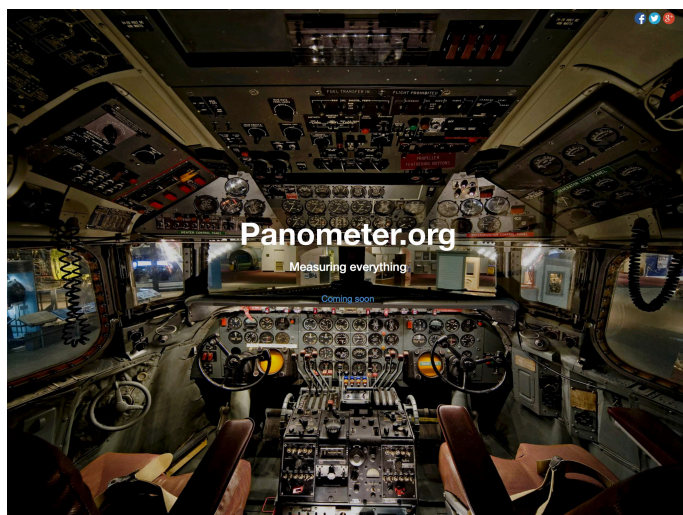Panometrics
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earthquake:

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earthquake:

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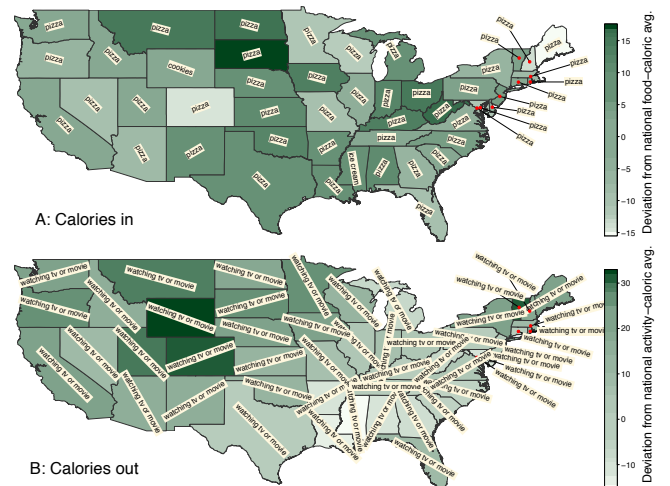@networksvox
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earthquake:

Images from Scott McCloud's "Making Comics." [28]

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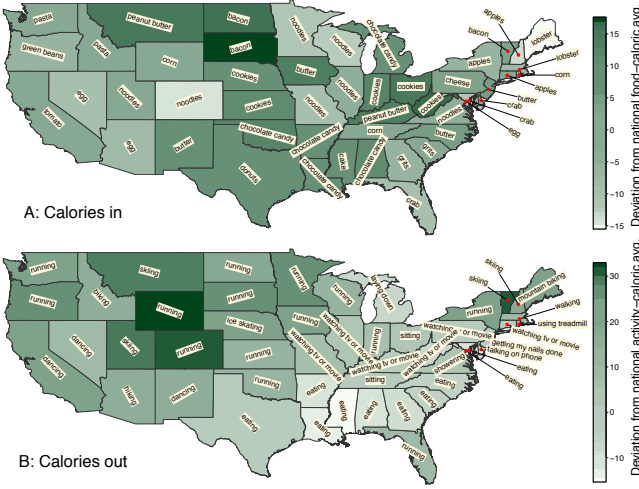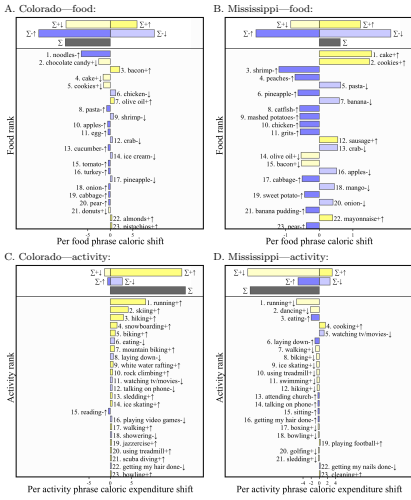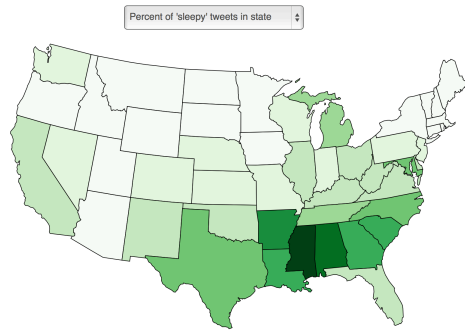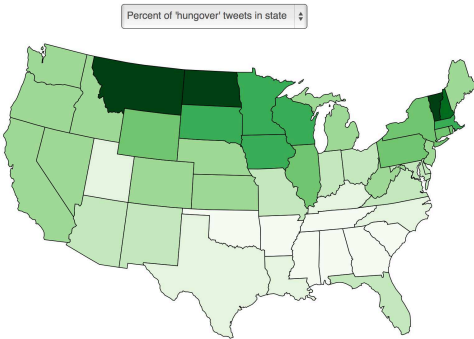Other Emotions
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laughter:

war:

rotting:

tortured:

lottery:

veterans:
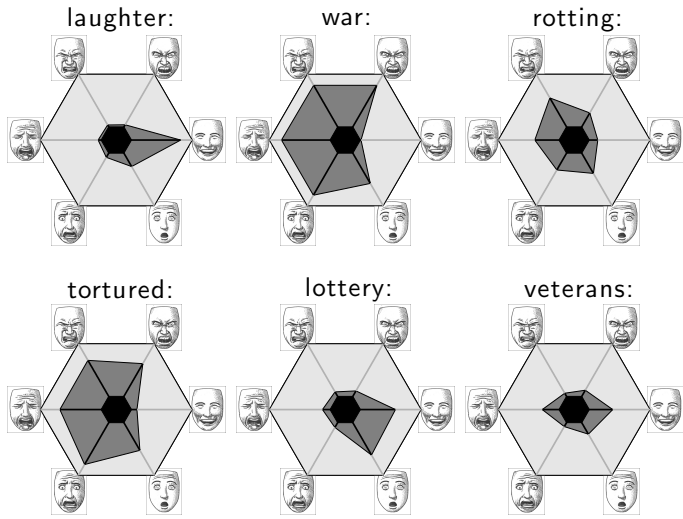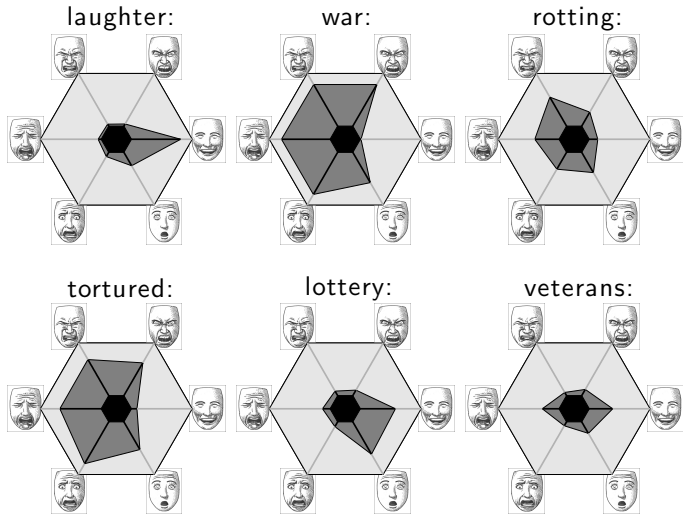
Word	ρ	p-value
cafe	-0.509	6.07×10^{-14}
sushi	-0.487	9.93×10^{-13}
brewery	-0.469	8.67×10^{-12}
restaurant	-0.448	8.93×10^{-11}
bar	-0.435	3.59×10^{-10}
banana	-0.434	3.77×10^{-10}
apple	-0.408	5.22×10^{-9}
fondue	-0.403	8.34×10^{-9}
wine	-0.400	1.08×10^{-8}
delicious	-0.392	2.17×10^{-8}
dinner	-0.386	3.85×10^{-8}
coffee	-0.384	4.51×10^{-8}
bakery	-0.383	5.12×10^{-8}
bean	-0.378	7.88×10^{-8}
espresso	-0.377	8.47×10^{-8}
cuisine	-0.376	8.82×10^{-8}
foods	-0.374	1.07×10^{-7}
tofu	-0.372	1.27×10^{-7}
brunch	-0.368	1.79×10^{-7}
veggie	-0.364	2.46×10^{-7}
organic	-0.361	3.13×10^{-7}
booze	-0.360	3.34×10^{-7}
grill	-0.354	5.4×10^{-7}
chocolate	-0.351	6.77×10^{-7}
#vegan	-0.350	7.47×10^{-7}

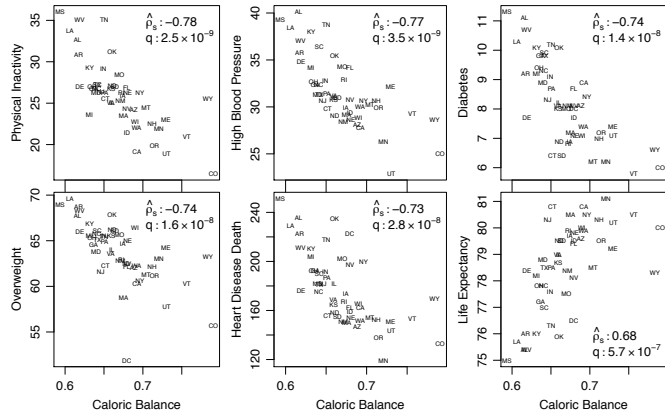|		
mcdonalds	0.246	6.18×10^{-4}
eat	0.241	8.22×10^{-4}
wings	0.222	2.13×10^{-3}
hungry	0.210	3.65×10^{-3}
heartburn	0.194	7.37×10^{-3}
ham	0.177	1.45×10^{-2}

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"The Lexicocalorimeter: Gauging public health through caloric input and output on social media"

Alajajian, Williams, Reagan, Alajajian, Frank, Mitchell, Lahne, Danforth, and Dodds. PLoS ONE, **12**, e0168893, 2017. [1]

http://panometer.org/instruments/lexicocalorimeter/

Panometer.org

Measuring everything

Coming soon

The Lexicocalorimeter:

A: Calories in

Deviation from national food-caloric avg.

B: Calories out

Deviation from national activity-caloric avg.

The Lexicocalorimeter:

A: Calories in

B: Calories out

Deviation from national food–caloric avg.

Deviation from national activity–caloric avg.

The Insomniometer:

Percent of 'sleepy' tweets in state

Correlation with CDC's Sleep Insufficiency measure:
$r_s = .48, p = 5.3 \times 10^{-4}$.

The Lexicocalorimeter:

A. Colorado—food:

B. Mississippi—food:

C. Colorado—activity:

D. Mississippi—activity:

Per food phrase caloric shift

Per activity phrase caloric expenditure shift

The Hangoverometer:

Percent of 'hungover' tweets in state

Correlation with Binge drinking estimates: $r_s = .72$, $p = 5.8 \times 10^{-9}$.

The Lexicocalorimeter:

Physical Inactivity $\hat{\rho}_s : -0.78$ $q : 2.5 \times 10^{-9}$

High Blood Pressure $\hat{\rho}_s : -0.77$ $q : 3.5 \times 10^{-9}$

Diabetes $\hat{\rho}_s : -0.74$ $q : 1.4 \times 10^{-8}$

Overweight $\hat{\rho}_s : -0.74$ $q : 1.6 \times 10^{-8}$

Heart Disease Death $\hat{\rho}_s : -0.73$ $q : 2.8 \times 10^{-8}$

Life Expectancy $\hat{\rho}_s : 0.68$ $q : 5.7 \times 10^{-7}$

Caloric Balance

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