

# Tutorial outline

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## ◆ Core sentiment analysis (SA) methods

- **Simple:** using lexica (dictionaries)
- **Aspect-based:** using information extraction

## ◆ Machine learning for SA

- Unsupervised: open language SA (LDA)
- Supervised: regression and deep learning

## ◆ SA extensions

- Generalized sentiment: person and community
- Multilingual, multimedia



# Generalized sentiment: emotion, personality, and subjective well-being

Johannes Eichstaedt, Peggy Kern, Marty Seligman,  
Andy Schwartz

**Lyle Ungar**

# Communication is more than facts

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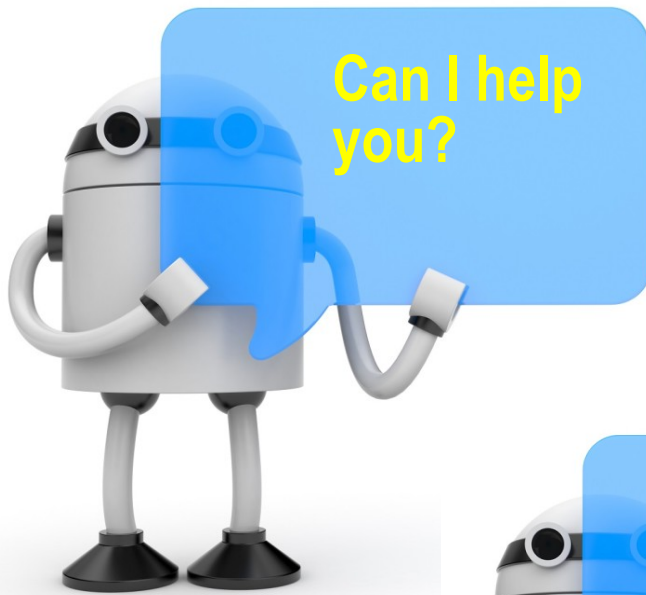


Photo credit: <https://onlinelearninginsights.wordpress.com/2016/06/07/need-to-know-news-chatbots-the-new-online-teaching-assistant-and-credit-worthy-moocs-go-global/>

# Sentiment beyond like/dislike

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## ◆ What people feel

- Emotion
- Stress
- Empathy

States

## ◆ Who people are

- Personality

Traits

## ◆ What communities think and feel



# Emotion

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- ◆ anger,
- ◆ fear,
- ◆ disgust,
- ◆ happiness,
- ◆ sadness,
- ◆ surprise

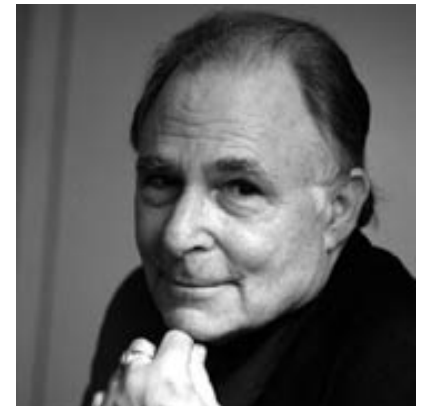
[http://www.cio.com/article/5\\_facial-expressions-test](http://www.cio.com/article/5_facial-expressions-test)

# Other universal emotions

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- ◆ Amusement
- ◆ Contempt
- ◆ Contentment
- ◆ Embarrassment
- ◆ Excitement
- ◆ Guilt
- ◆ Pride in achievement
- ◆ Relief
- ◆ Satisfaction
- ◆ Sensory pleasure
- ◆ Shame

Paul Ekman



# Use hashtags as labels

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## ◆ “distant supervision”

- 
1. *Feeling left out... #sadness*
  2. *My amazing memory saves the day again! #joy*
  3. *Some jerk stole my photo on tumblr. #anger*
  4. *Mika used my photo on tumblr. #anger*
  5. *School is very boring today :/ #joy*
  6. *to me.... YOU are ur only #fear*
- 

Using Hashtags to Capture Fine  
Emotion Categories from Tweets

# Personality – 5 Factor Model

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- ◆ **extroversion vs. introversion**
  - sociable, assertive vs. aloof, shy
- ◆ **neuroticism vs. emotional stability**
  - insecure, anxious vs. calm, unemotional
- ◆ **agreeableness (high vs. low)**
  - friendly, cooperative vs. antagonistic, fault-finding
- ◆ **conscientiousness (high vs. low)**
  - self-disciplined, organized vs. inefficient, careless
- ◆ **openness to experience vs. conventionality**
  - intellectual, insightful vs. shallow, unimaginative

# Emotion correlates with personality

EXT	NEU	AGR	CON	OPN
#unimportant	#happiness	#like	#calm	#tranquil
#attached	#bugged	#crushed	#considerate	#lust
#destroyed	#alert	#mixed	#mean	#love
#detached	#irate	#delighted	#cowardly	#appreciative
#awful	#cold	#happiness	#bewildered	#peaceful
#lust	#anxious	#hopeful	#shunned	#jealousy
#comfortable	#weak	#blushing	#exposed	#innerpeace
#burned	#shame	#jealousy	#imperfect	#thoughtful
#troubled	#sexy	#grateful	#judged	#touched
#jumpy	#desire	#spirited	#pity	#careful

As do words, images, Facebook likes ...

# Or get people to label words

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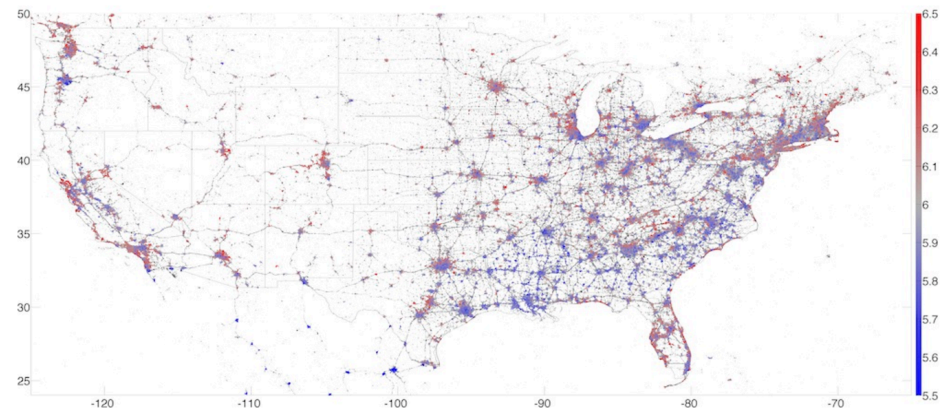
## ◆ Linguistic Inquiry and Word Count (LIWC)

- POSEMO, NEGEMO ....

## ◆ WordNet Affect

## ◆ NRC Emotion Lexicon

Dodds hedonometer



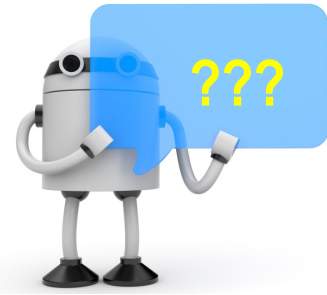
# Like most SA, a different use of NLP

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Historically:

◆ *what* people say

- Information extraction, question answering, ...



Today:

◆ *how* people say it

- Language variation with gender, age, personality, education, mood, health, stress, optimism, empathy, depression, satisfaction with life

# Closed vs. Open Language

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- ◆ Closed: use a lexicon
- ◆ Open: use words, LDA topics



# **Person-level models require different data collection**

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- ◆ **Not just using people to generate labels**
- ◆ **But learning about people**
  - People share their social media and
    - take questionnaires
    - or share their medical record
    - or shopping history, ...

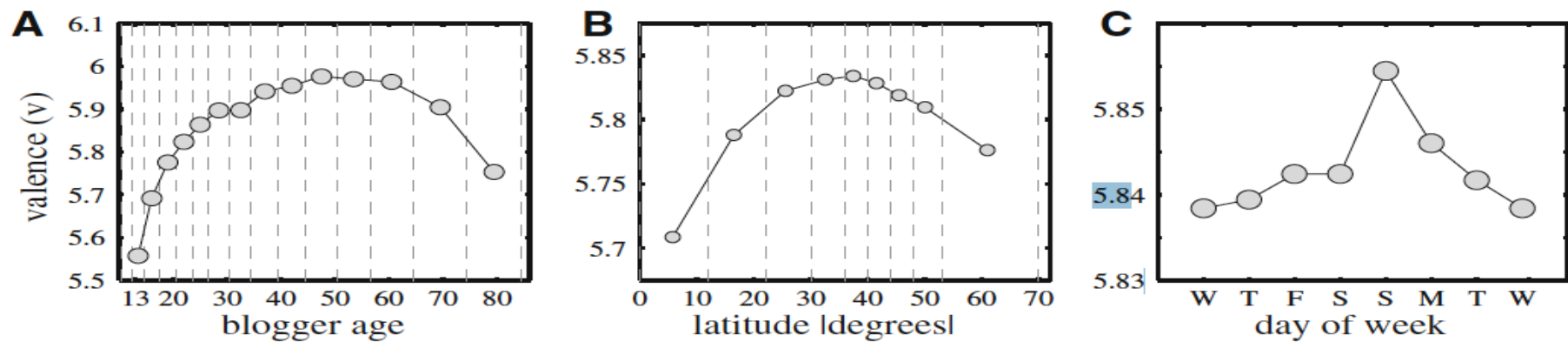
# LIWC happiness lexicon

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- accept
- accepta\*
- accepts
- advantag\*
- agreeing
- ....
- happy
- ...
- merry
- ....
- heartfelt
- heaven\*
- heh\*
- helper\*
- helps
- hilarious
- hoho\*
- honest\*
- honour\*
- hoped
- hopefully
- hopes
- terrific\*
- thank
- thanked
- thanks
- toleran\*
- treasur\*
- treat
- trueness
- truer
- truest
- truly
- trust\*
- values
- valuing

Pennebaker

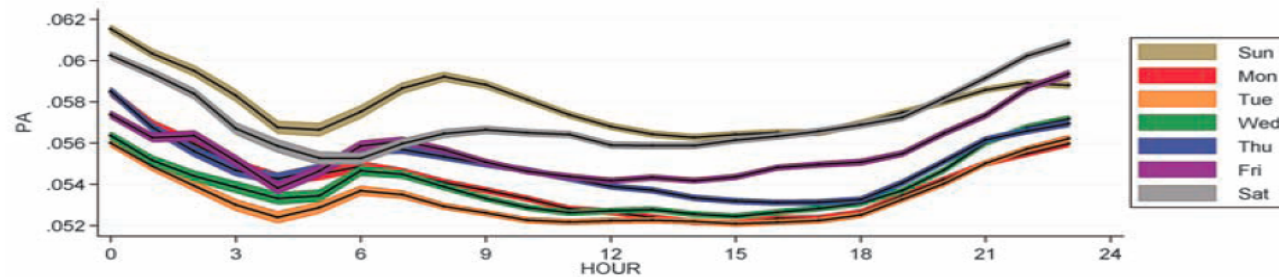
# Blogger Happiness



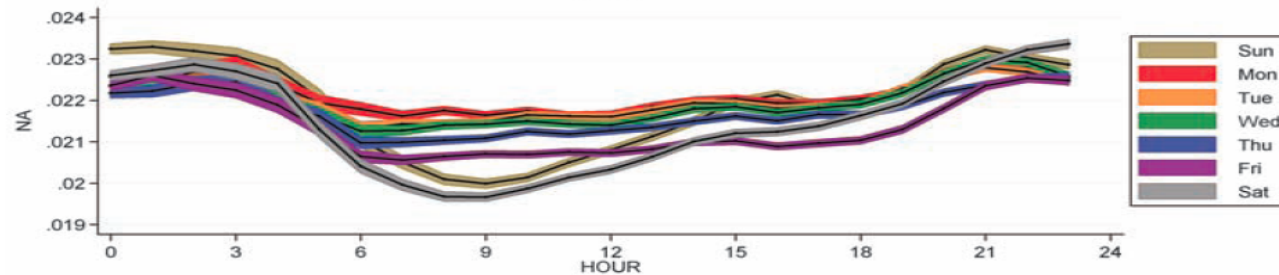
Measuring the Happiness of Large-Scale Written  
Expression: Songs, Blogs, and Presidents  
Dodds and Danforth 2010

# Twitter Happiness

Positive  
Affect



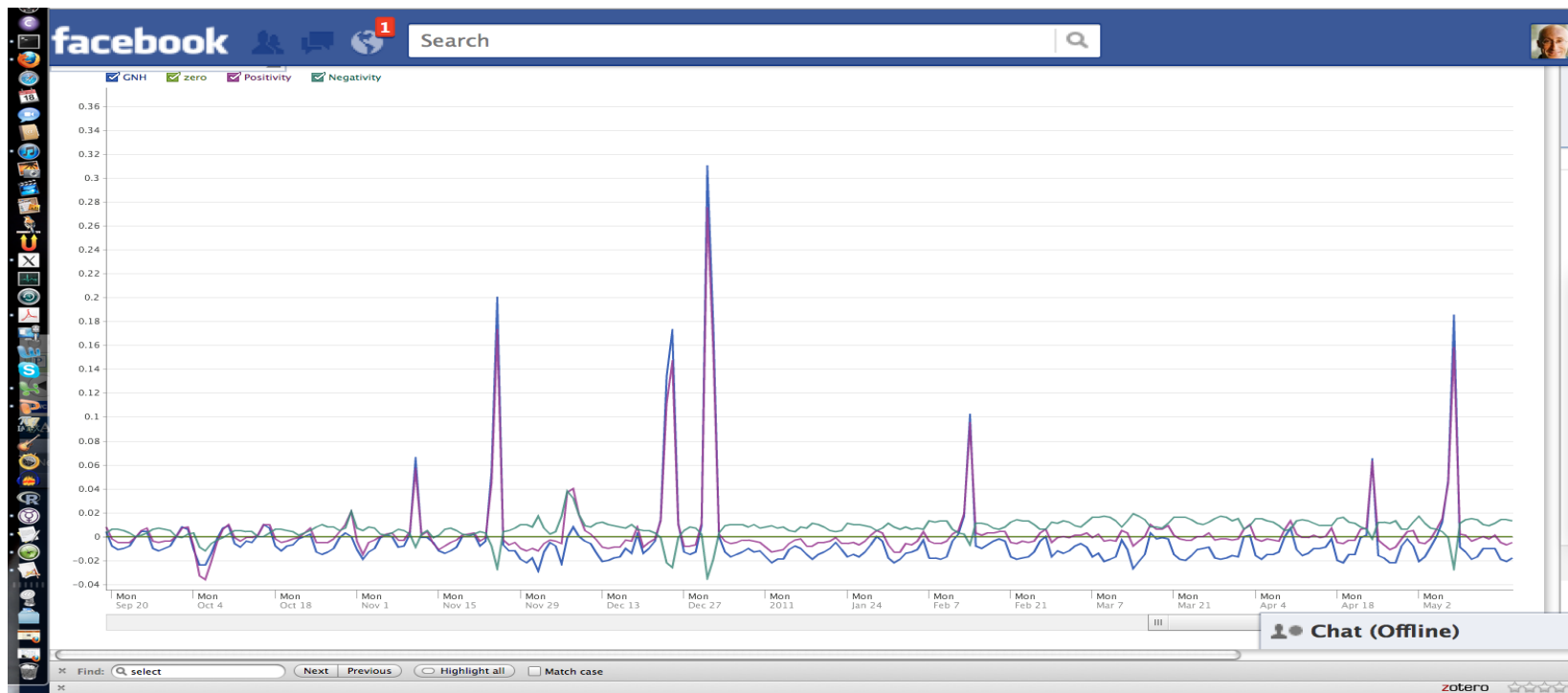
Negative  
Affect



Hour of day

Diurnal and Seasonal Mood Vary with Work, Sleep, and Daylength Across Diverse Cultures  
Golder\* and Macy (2011)

# Facebook Gross National Happiness



Kramer, A. D. I. (2010). An unobtrusive model of "gross national happiness." Proc. CHI, 2010, ACM Press, 287-290.

# Facebook Gross National Happiness

What are they measuring?



Kramer, A. D. I. (2010). An unobtrusive model of "gross national happiness." Proc. CHI, 2010, ACM Press, 287-290.

# Be careful about ambiguity

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- ◆ *Happy birthday*
- ◆ *Merry Christmas*
- ◆ *Michael Jackson is dead*
- ◆ *Great Britain*
- ◆ *Legal tender*

Lexicon-based results can be flawed

# Open-vocabulary personality estimation from Facebook

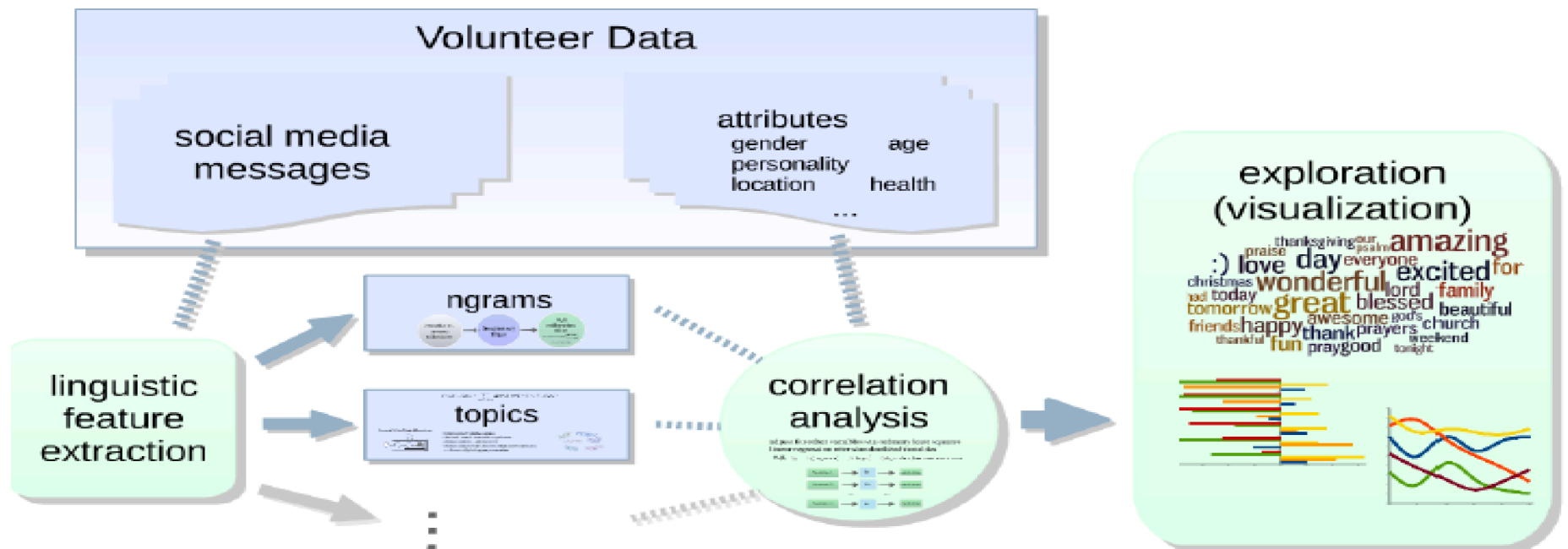
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- ◆ Use Facebook posts, demographics and personality tests from 70,000 people
  - To find words that most correlate with sex, age, IQ, happiness, personality

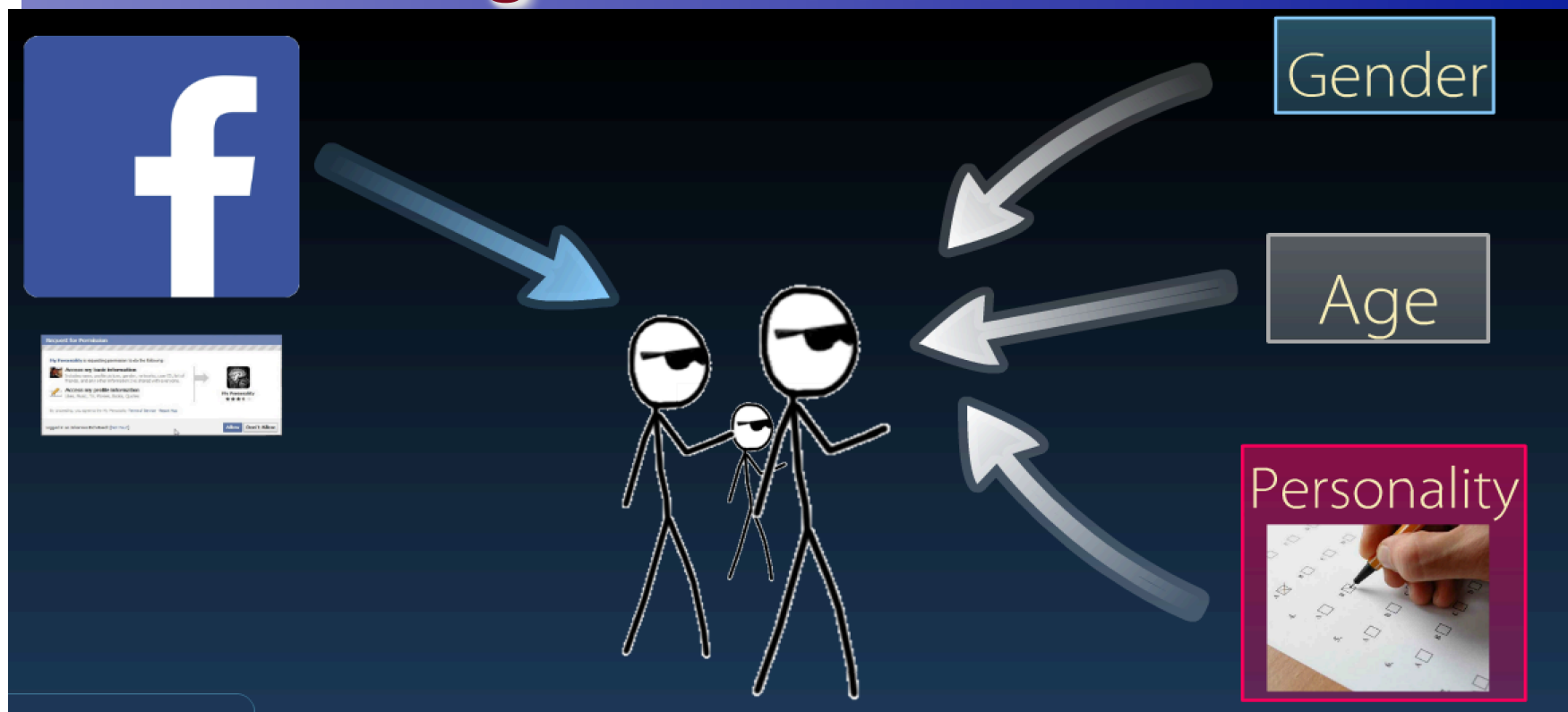
[wwbp.org](http://wwbp.org)



# Open Vocabulary Analysis Method



# Modeling Individuals



**N=70,000**

# Females



[illegible]

# Females







Gender

Accuracy:  
92%







# Intraversion



# Neurotic words



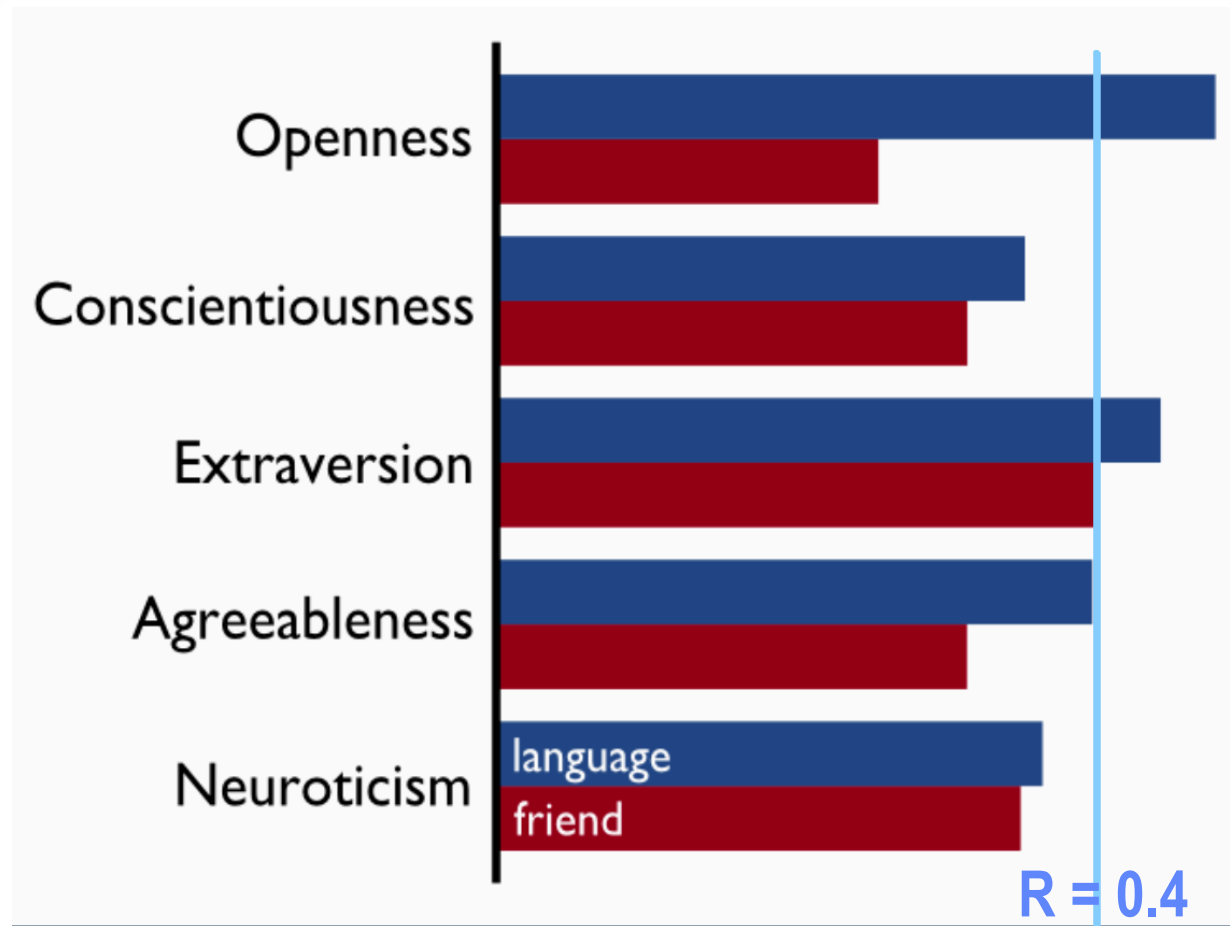
# Well-adjusted (non-neurotic) words





# Language vs. Friends

Language predicts  
Personality as well  
As friends do.



# Depression



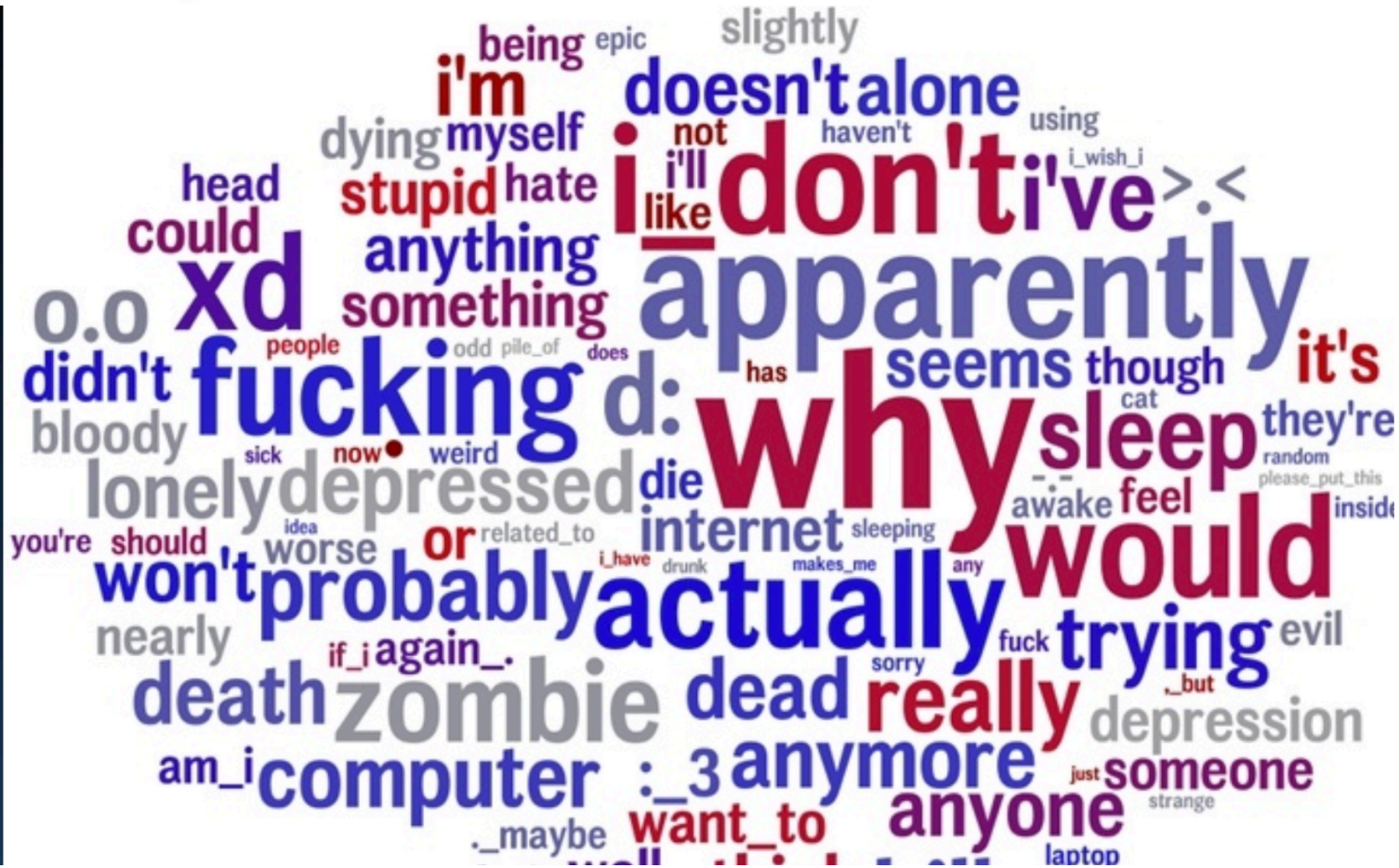


# Depression – low mood

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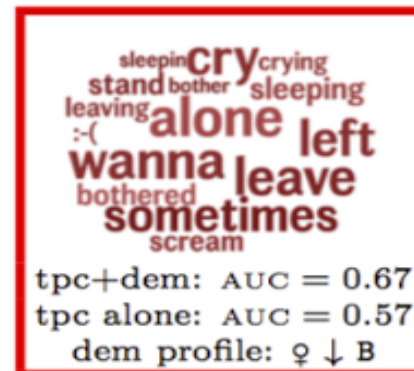
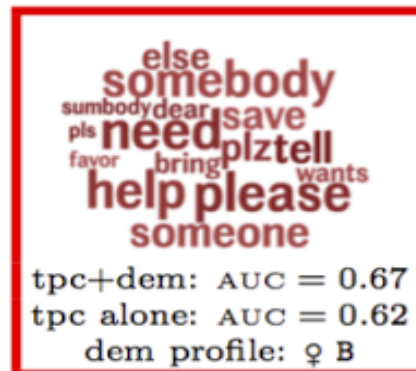
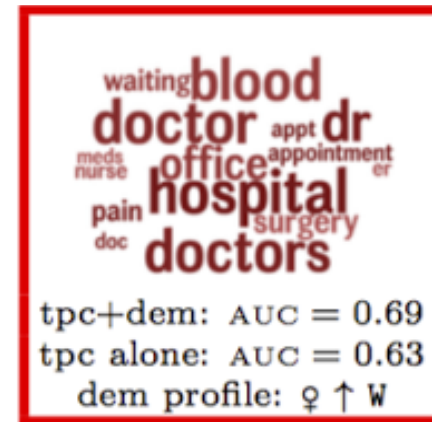
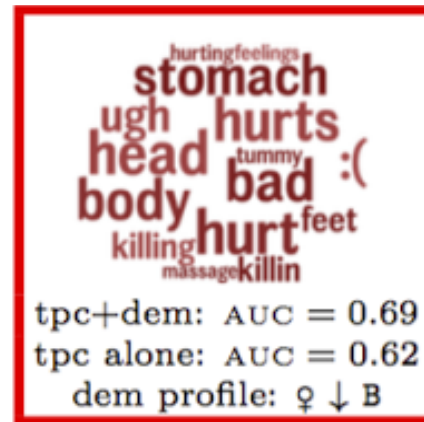
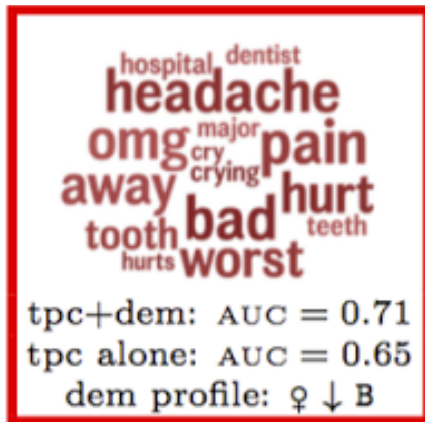


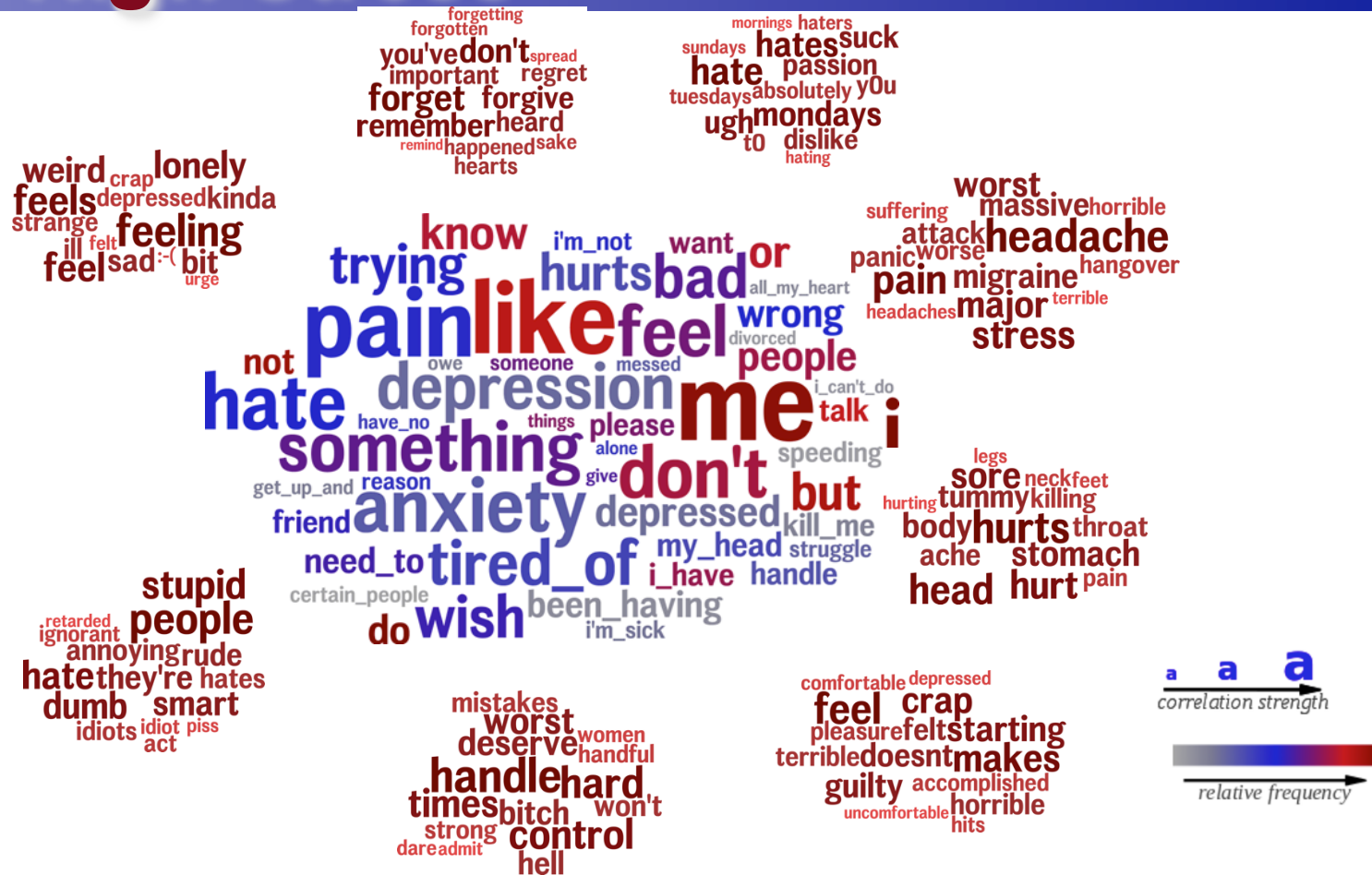
# Depression – low self-worth

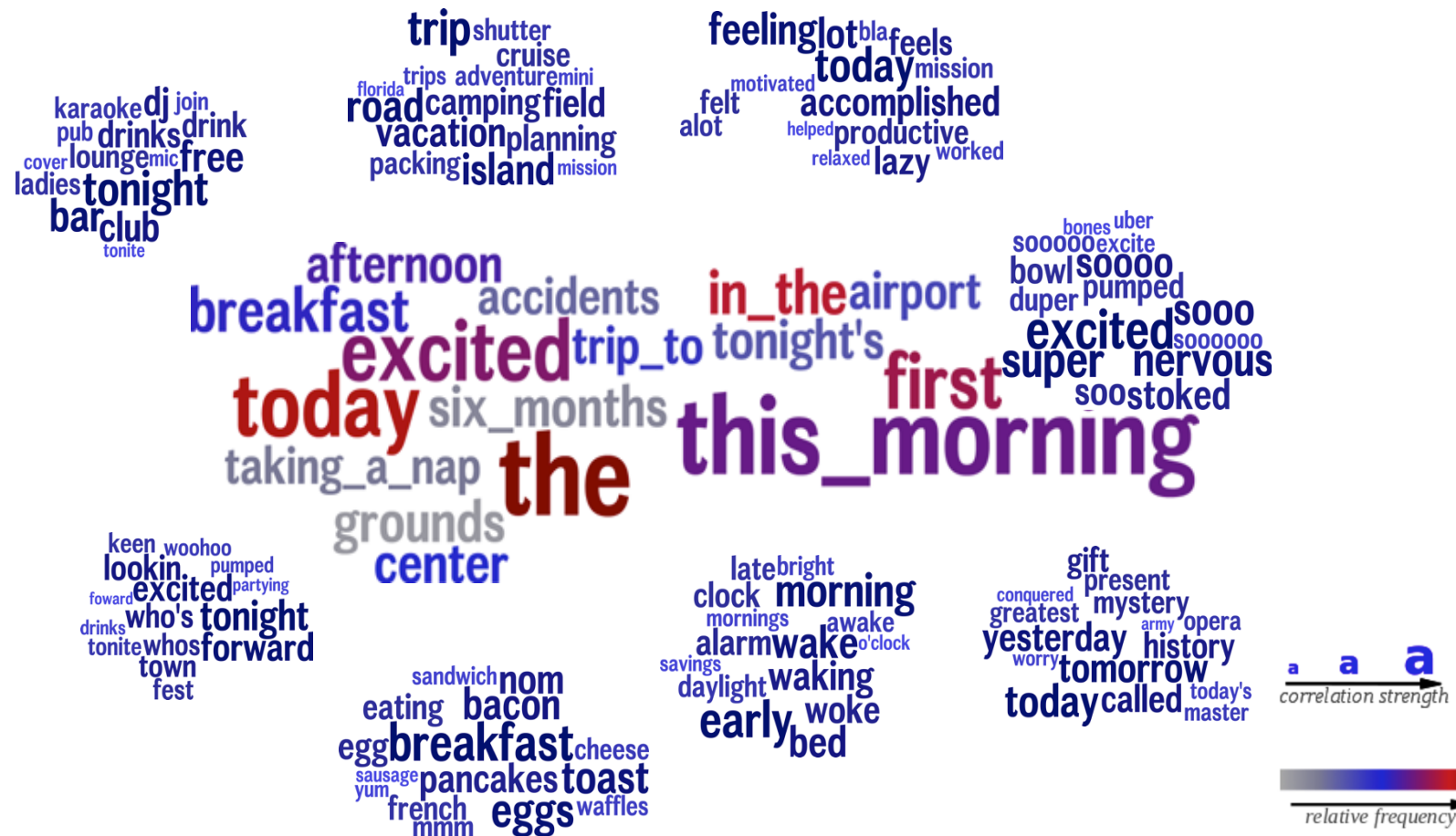




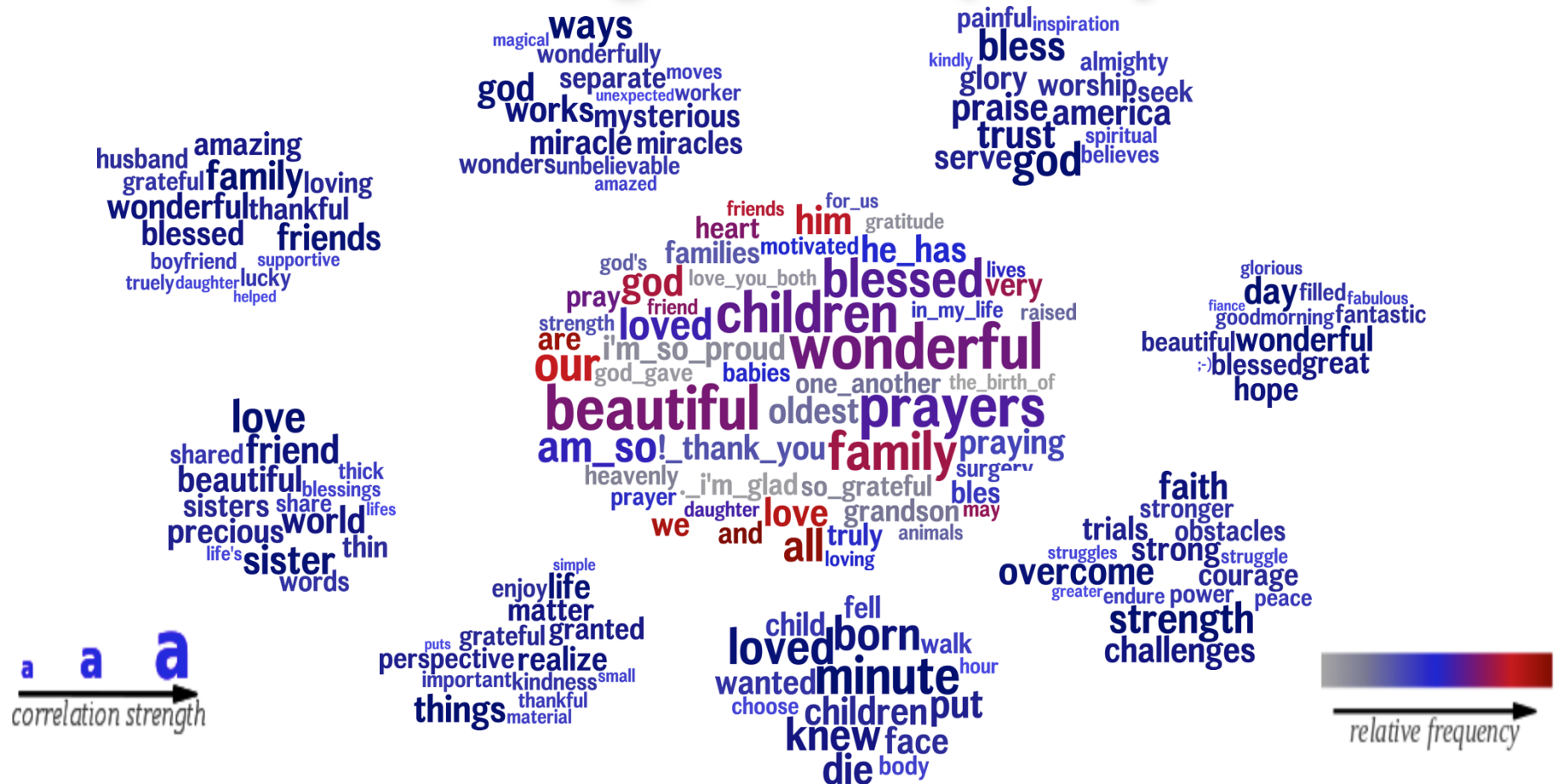
# Depression (ER sample)



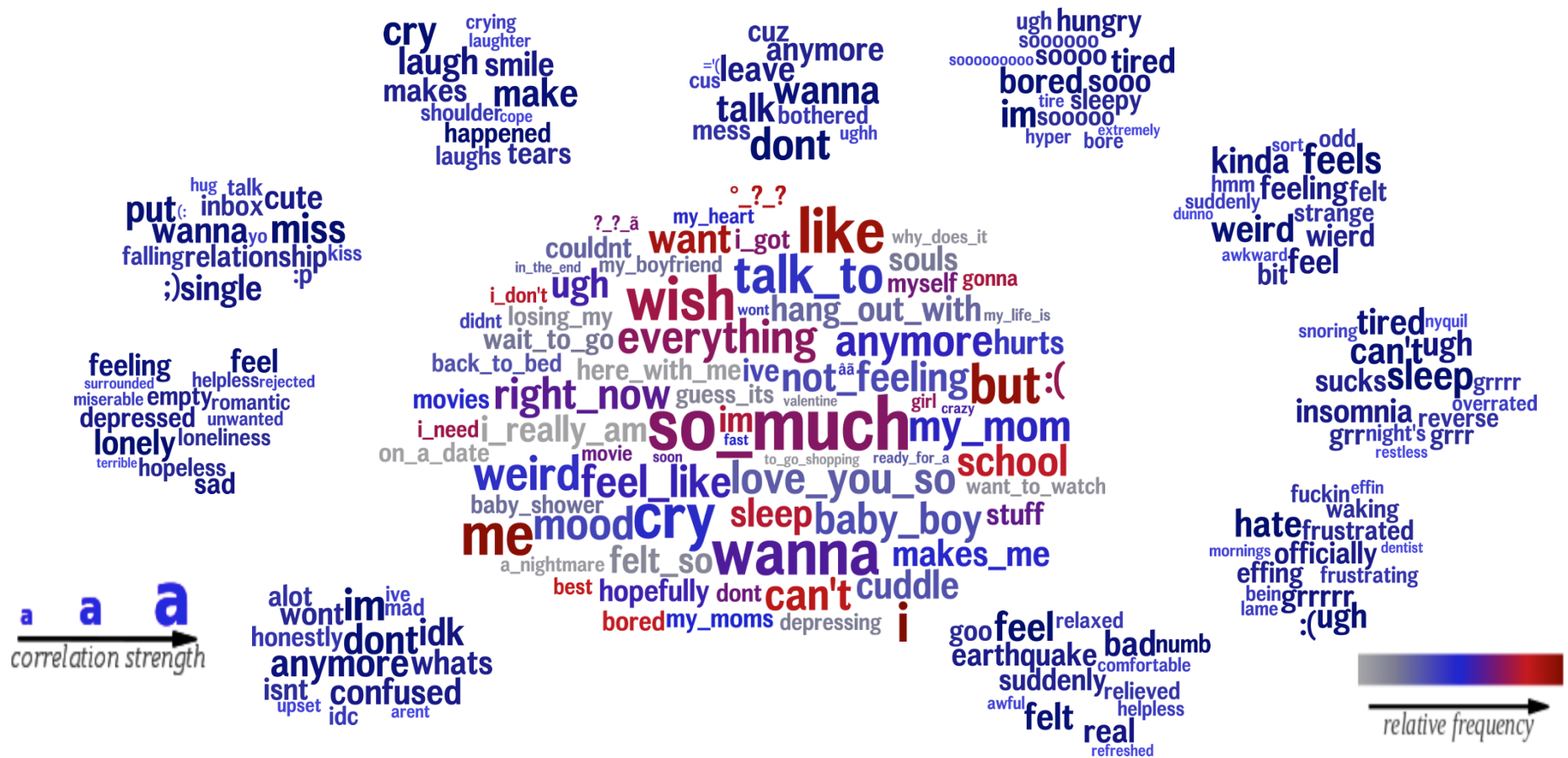




# Salutogenic empathy



# Pathogenic empathy



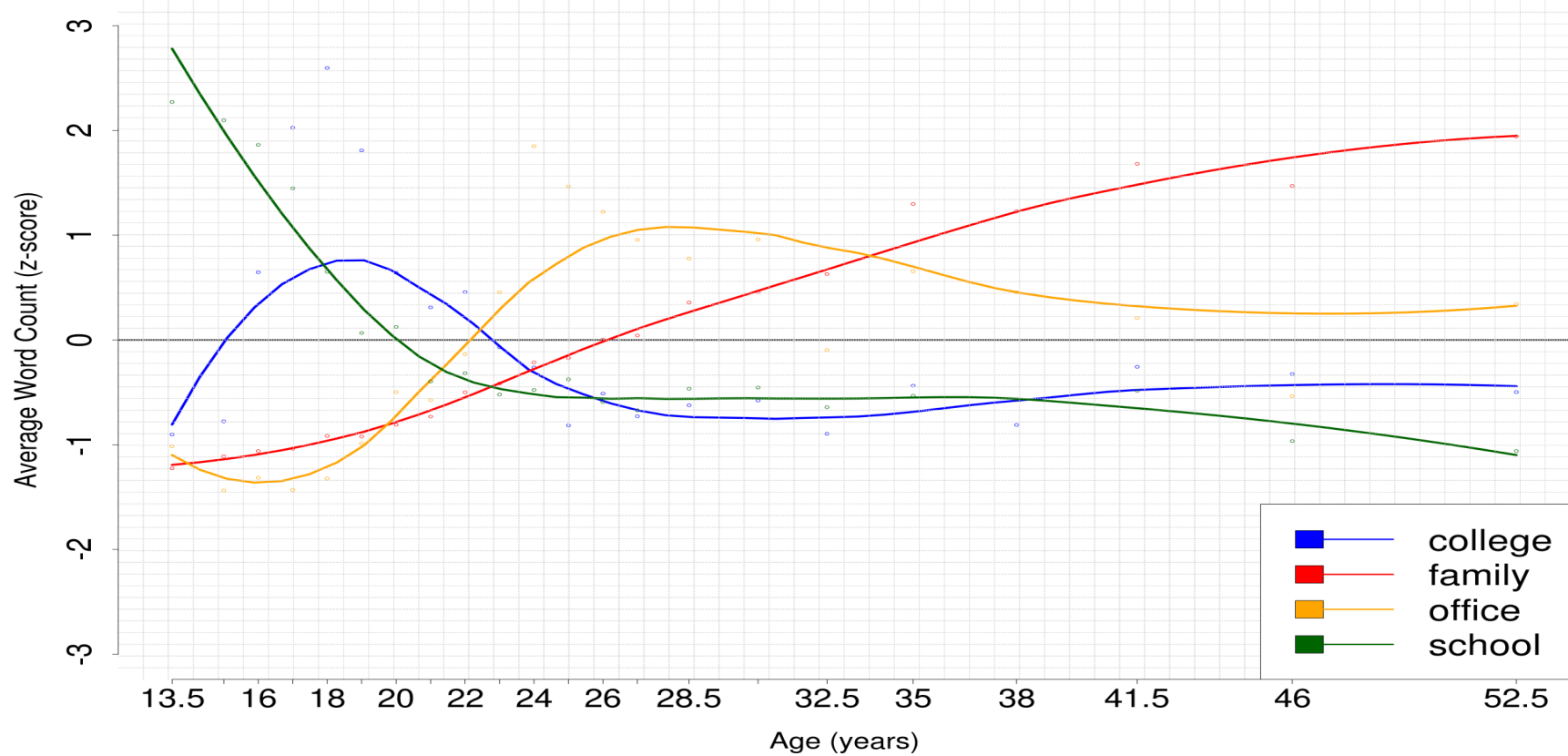
# Visualization is important

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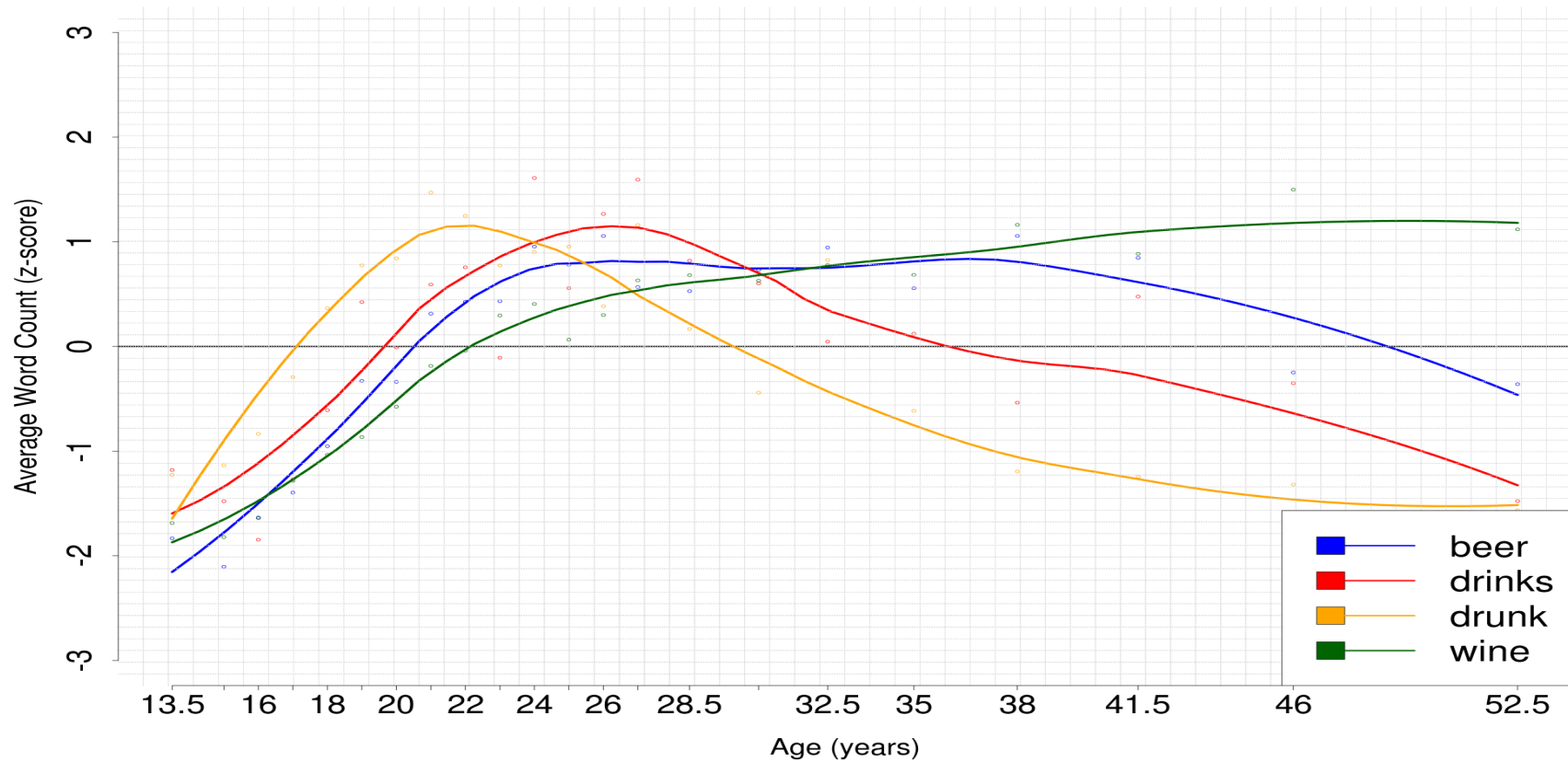
- ◆ **Changes in sentiment over time**
- ◆ **Differences between populations by**
  - Demography etc.
    - Age, sex, country of origin
    - Income, education, political orientation
  - Location
  - Time of day
  - Device
  - ...



# Life stages

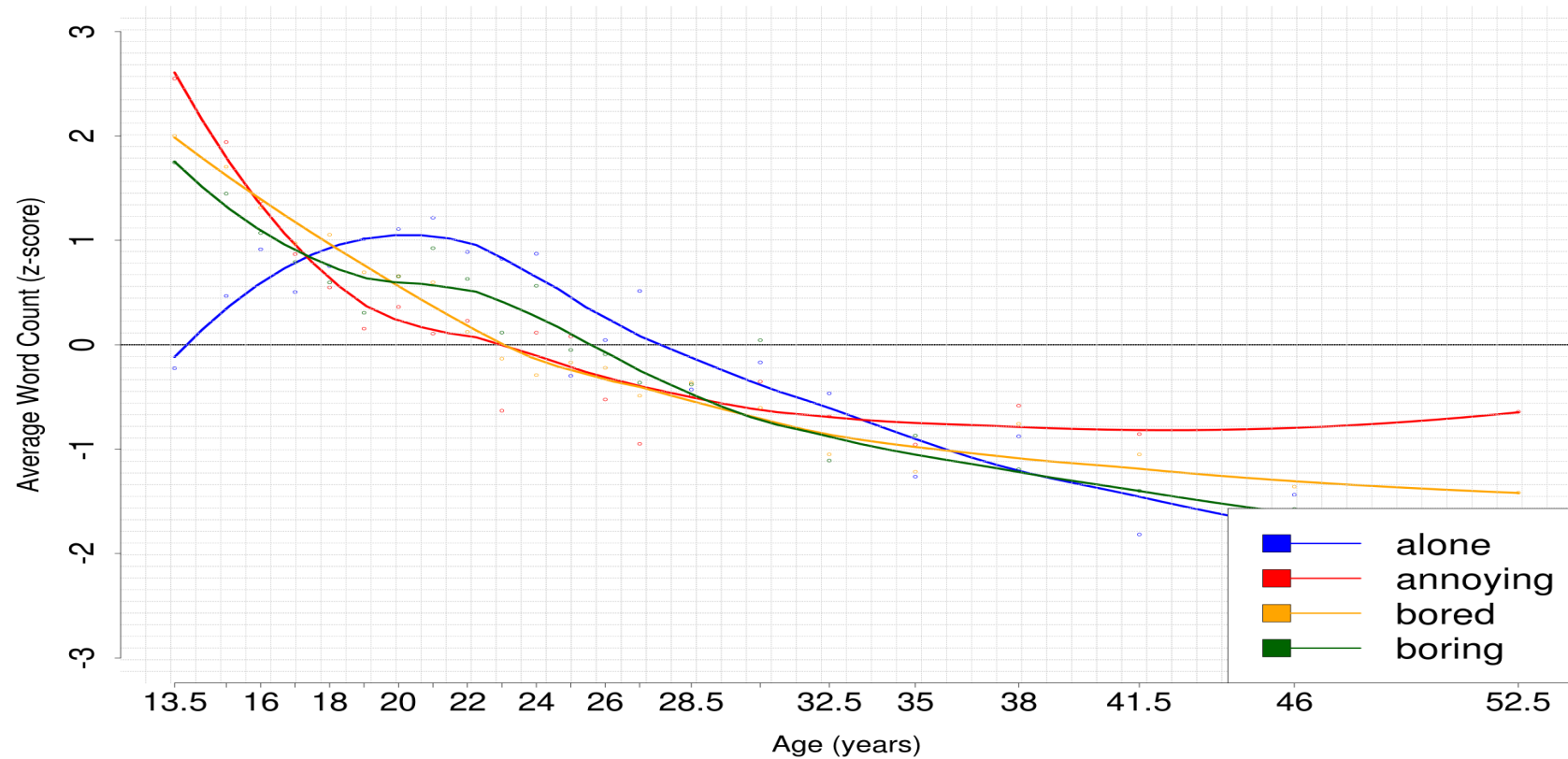


# Alcohol consumption

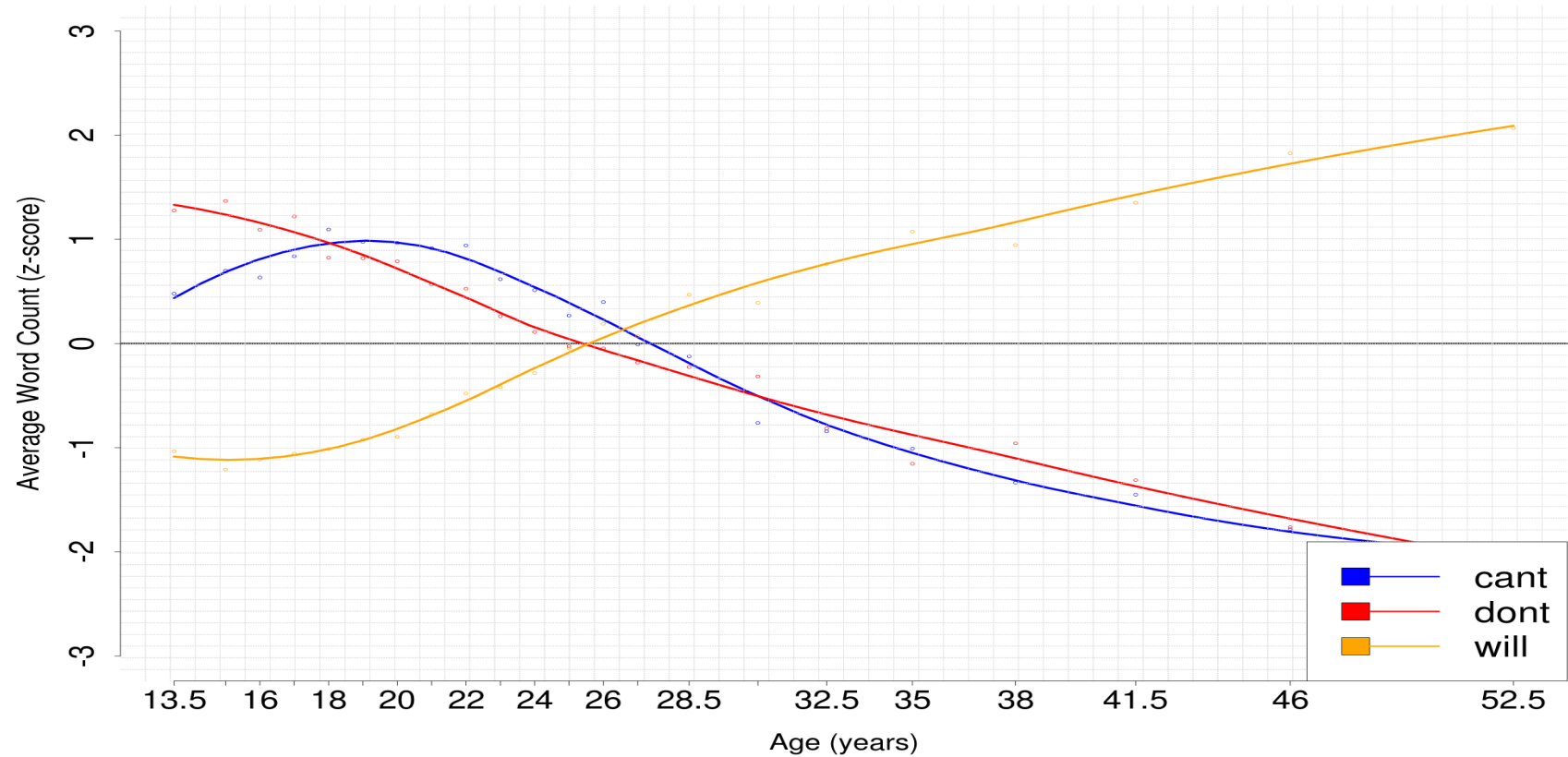




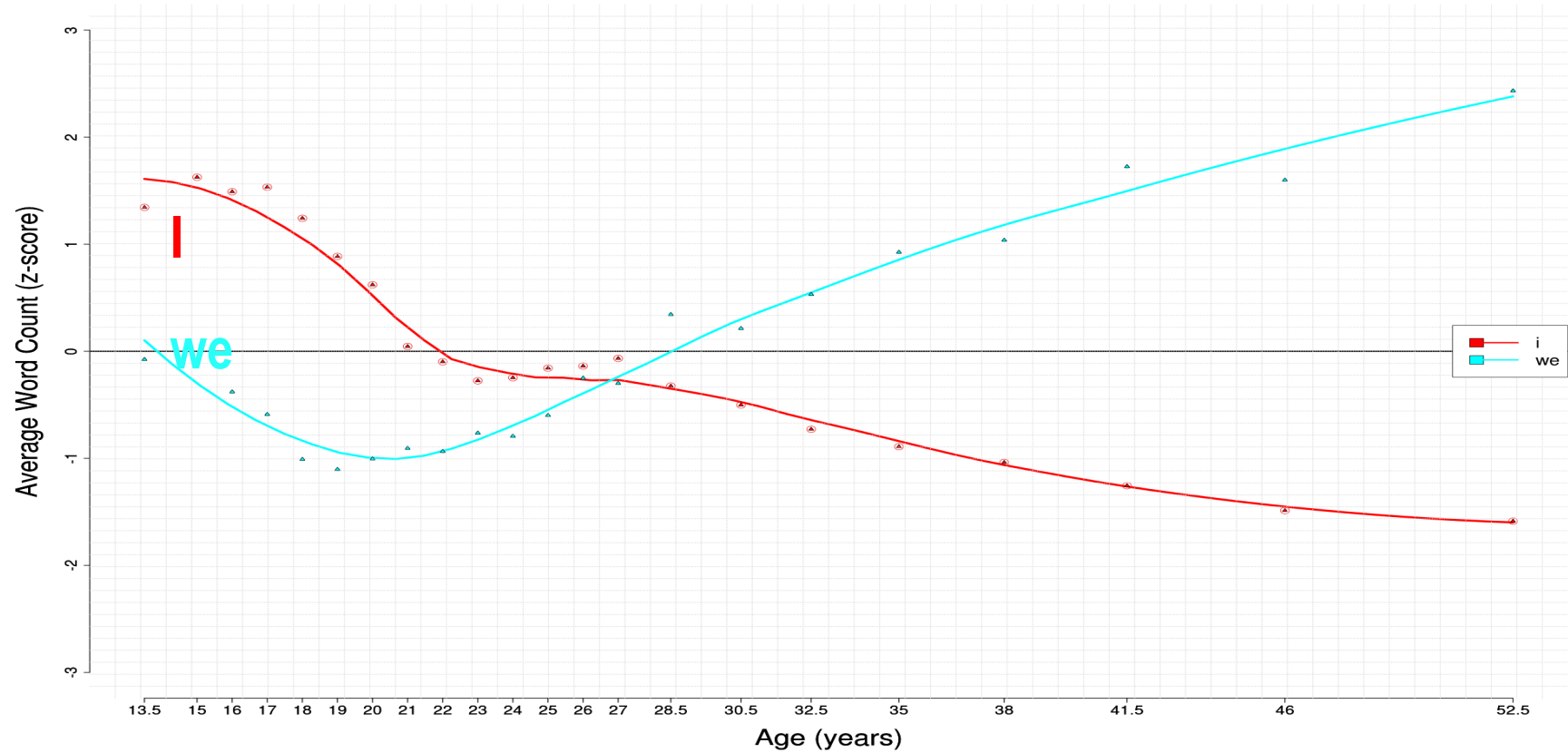
# Older people are less negative



# Older people are more positive



# Older people are more other-oriented



# Person-level Takeaways

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- ◆ **Language reveals personality, conveys emotions**
  - Images do too!
- ◆ **Crowdsourcing lets us predict personal traits from language**
  - Age, sex, race
  - Personality
  - Stress, burnout, emotion, life satisfaction
  - Political beliefs
  - Mental and physical health

# The World Well Being Project

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A word cloud featuring the names of individuals associated with The World Well Being Project. The names are arranged in a dense, overlapping cluster, with colors ranging from blue to red. The font sizes vary, with larger names indicating higher frequency or prominence. The names include: Yi Chen, Greta Jawel, Zahra Riahi Samani, Amanda Geiser, David Yaden, Sharath Chandra Guntuku, Youngseo Son, Peggy Kern, Rob DeRubeis, Brenda Curtis, Jordan Carpenter, Christie Versagli, Patrick Crutchley, Laura Smith, Courtney Hagan, Robert Backer, Lyle Ungar, Molly Ireland, Andy Schwartz, Jennifer Kregor, Han Lin, Martin Seligman, Hao Peng, Zhi Soon, Sal Giorgi, George E. Vaillant, David Stilwell, Leqi Liu, Scott Barry Kaufman, Eduardo Blanco, Louis Tay, Shawndra Hill, Johannes Eichstaedt, Daniel Preotiuc-Pietro, Sphoorti Mirji, Nick Fausti, Dan Rieman, Peter Schwartz, Anneke Buffone, Matthias Mehl, Saif Mohammed, Muhammad Abdul-Mageed, Ye Liu, Masoud Rouhizadeh, Jeanette Elstein, Michael Kosinski, Phillip Lu, Dean Fulgoni, Chris Tufts, João Sedoc, Raina Merchant, Tianli Han.

[wwbp.org](http://wwbp.org)