



JOHNS HOPKINS

WHITING SCHOOL  
*of* ENGINEERING

# Affect and Lexicons

# Recap

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- Word embeddings as methodology for corpus analysis
  - We often use embeddings to compute relations between sets of words:
    - {Woman, she, her, gal, girl}
    - {nurse, secretary, teacher}
  - Dimensions of beliefs
    - Gender, potency (power)
- Where do these words come from? What other types of word annotations are useful?

# This class

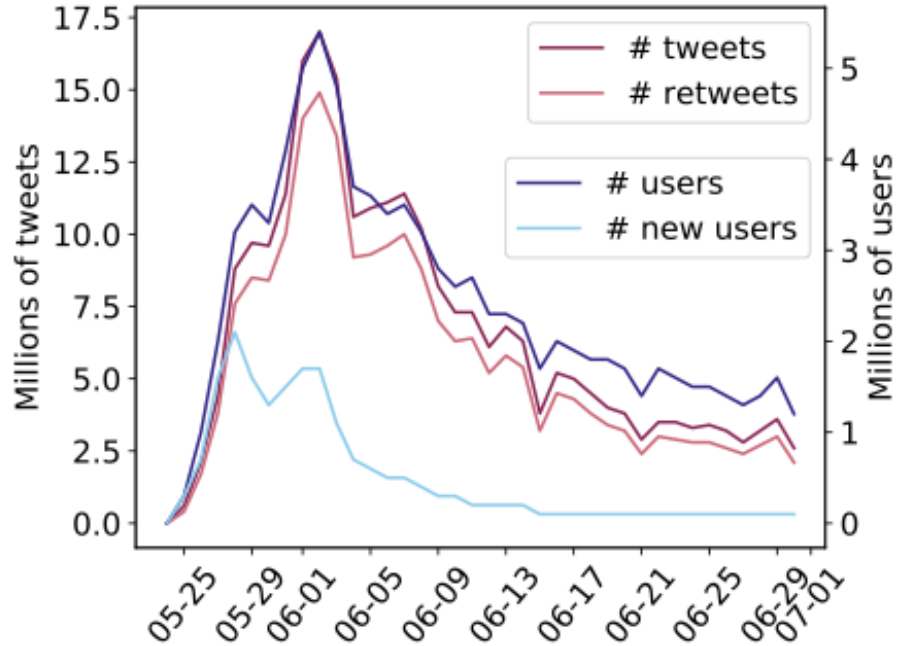
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- Psychology measures of affect and emotion
- Common lexicons, construction and uses
- [Data annotation]

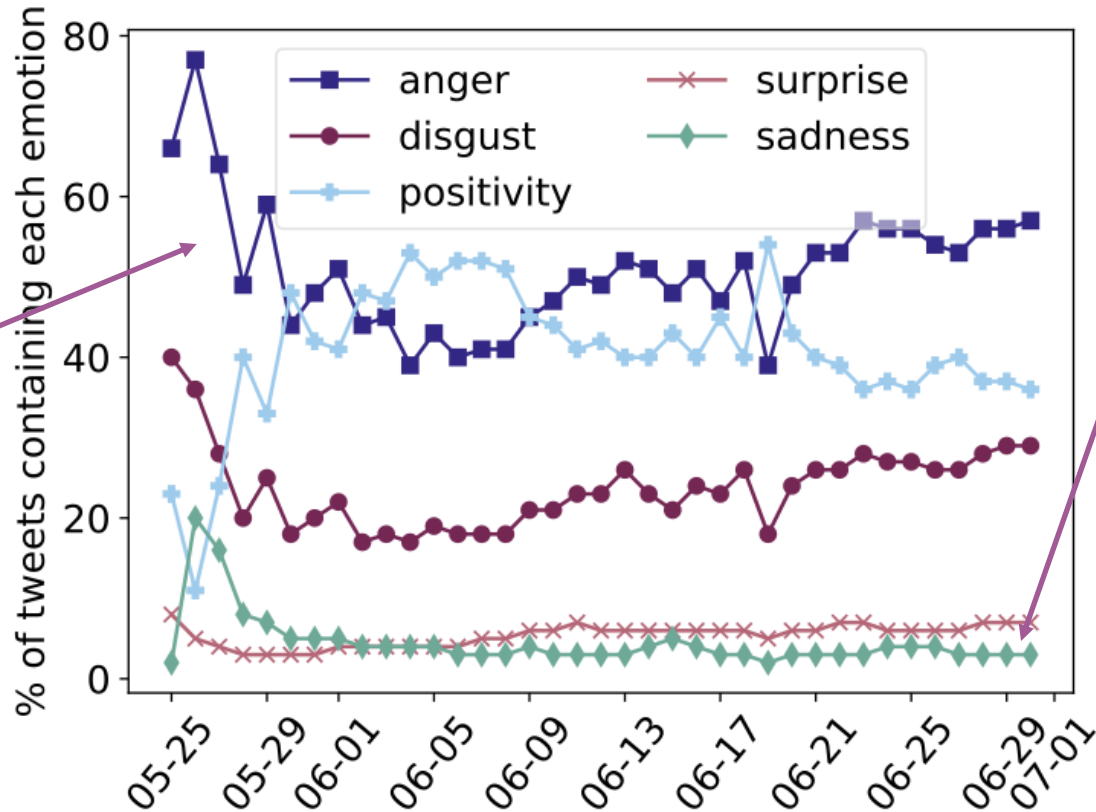
# Analysis Data: 34M tweets about the #BlackLivesMatter Movement

The term #BlackLivesMatter originated in posts made by activists Alicia Garza and Patrisse Cullors in 2013

#BlackLivesMatter  
#JusticeForGeorgeFloyd  
#ICantBreathe



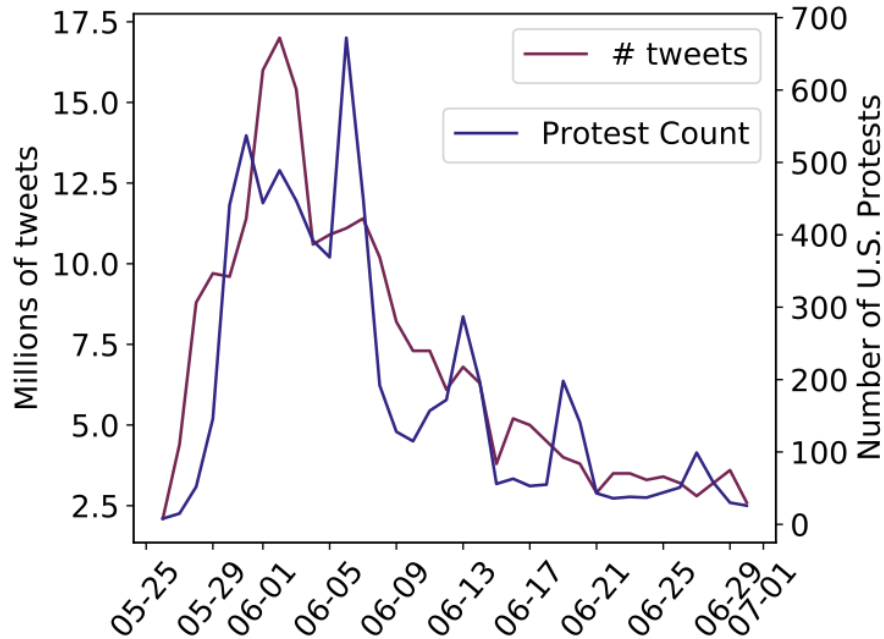
# Emotions over time in tweets with pro- BLM hashtags



Initial peak in anger, eclipsed by more positive emotions

Low sadness and surprise?

# Positivity is correlated with in-person protests



	Correlation with protest across states	Correlation with protests across cities
<b>Anger</b>	-0.43*	-0.16*
<b>Disgust</b>	-0.24	-0.21*
<b>Positivity</b>	0.48*	0.12*
<b>Sadness</b>	-0.38*	0.06
<b>Surprise</b>	-0.25	0.09



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# Emotion Taxonomies

# What is an emotion?

- **Emotions** are a mix of
  - (1) physiological arousal (heart pounding)
  - (2) expressive behaviors (quicken pace),
  - (3) consciously experienced thoughts (is this a kidnapping?) and feelings (a sense of fear, and later joy)
- The puzzle for psychologists has been figuring out how these three pieces fit together



# James-Lange Theory

**Stimulus**  
(Growling Dog)



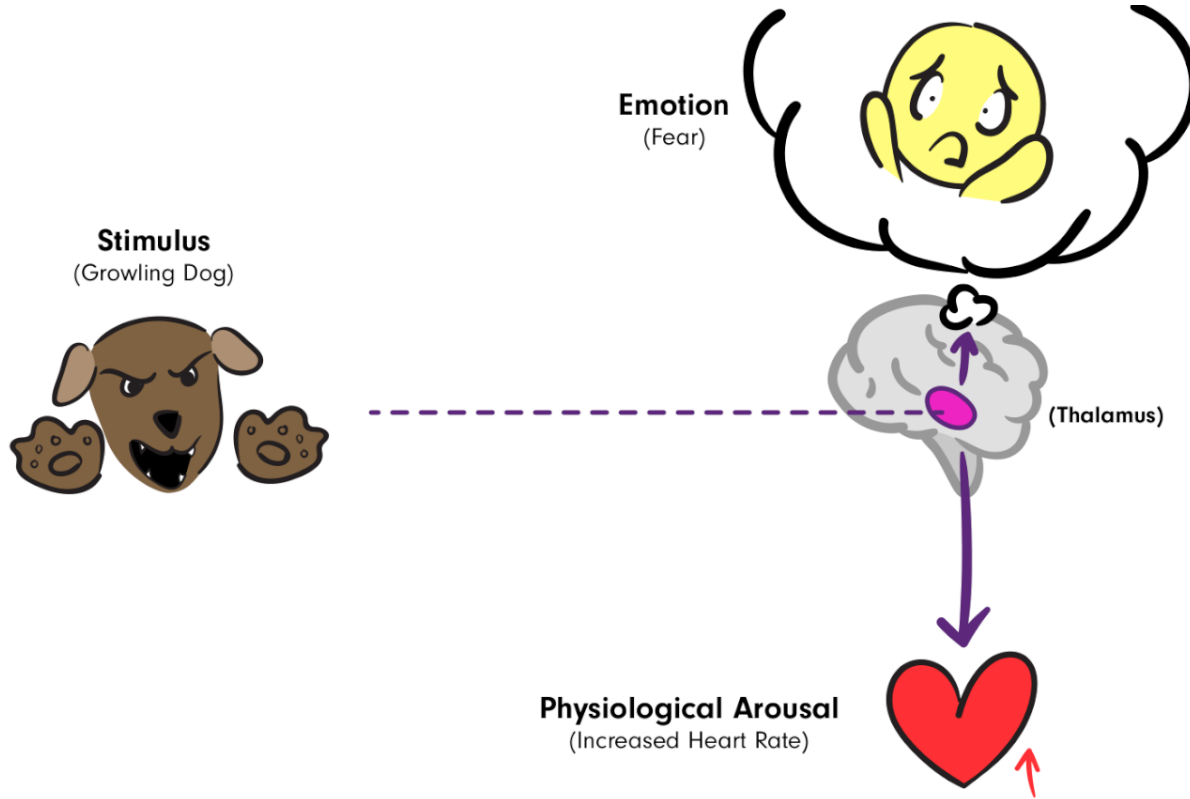
**Physiological Arousal**  
(Increased Heart Rate)



**Emotion**  
(Fear)



# Cannon-Bard Theory



# Discrete Emotion Theory

- All humans have innate set of basic emotions that are cross-culturally recognizable
- “Discrete”: emotions are separate and distinct
- Distinguishable by neural, physiological, behavioral and expressive features
- A little historical context:
  - Darwin (1872) described “several facial, physiological and behavioral processes associated with different emotions in humans as well as animals”
  - Tomkins (1962, 1963) proposed 8 “pancultural affect programs”: surprise, interest, joy, rage, fear, disgust, shame and anguish

# Paul Ekman and Carroll Izard Taxonomy

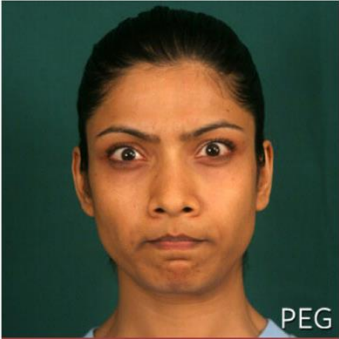
- “I and others found evidence...that certain facial expressions of emotion appeared to be universal”
- Example field experiments:
  - Show stress-inducing films to students in the US and Japan → Japanese and American students had virtually identical facial expressions
  - Show photographs of different emotion expressions to people in US, Japan, Chile, Argentina, and Brazil: people judged the same emotions in these countries
- Each basic emotion is a *family* of related states

# Paul Ekman's Taxonomy



- Sadness
- Anger
- Enjoyment
- Disgust
- Surprise
- Fear
- Contempt

# Paul Ekman's Taxonomy



Anger



Contempt



Disgust



Enjoyment



Fear



Sadness

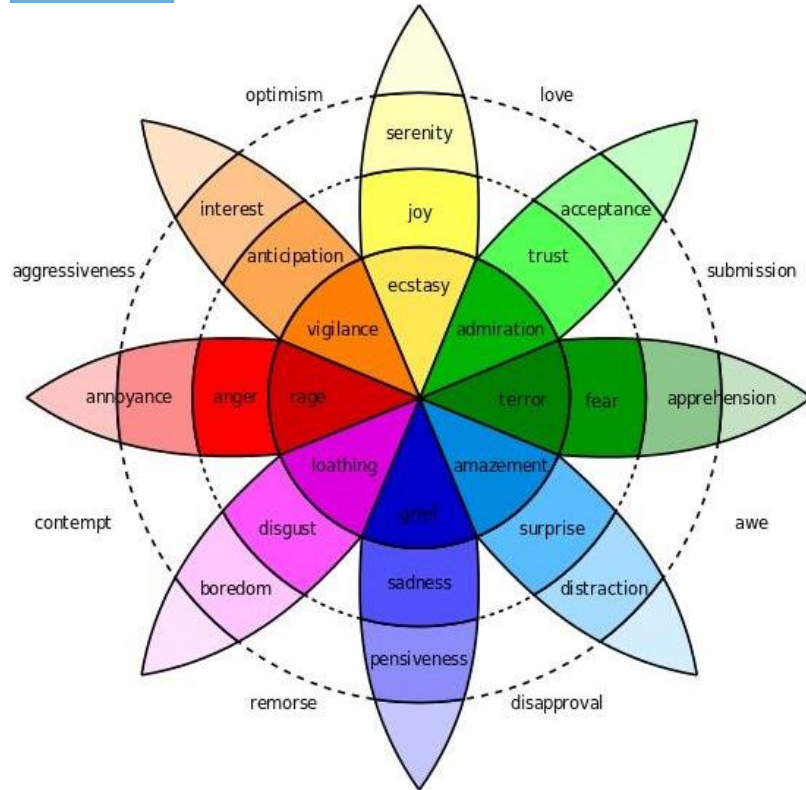


Surprise

# Critiques of Discrete Emotion Theory

- Failure to find correlations between neural and nervous system (ANS) activity and emotions
- Discrete Emotion Theory cannot account for rich variability and context-sensitivity of emotions (Russell and Barrett)
  - Factors other than immediate feeling can affect facial expressions (you may smile out of a desire to please others rather than happiness)
  - Emotions can elicit different responses: flight or fight response to fear
  - Expressions of emotions can differ across cultures

# Plutchik Emotion Taxonomy (Increasing continuity)



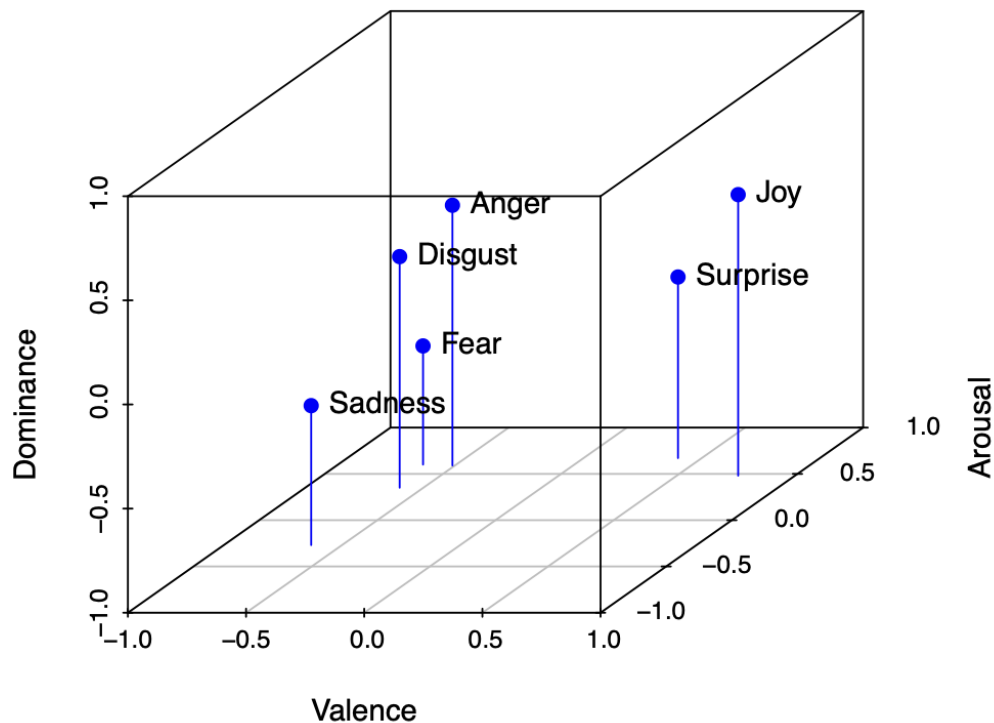
- Still have 8 basic emotions emotions in the center
- Different levels of intensity
- Some emotions are combinations of 8 core emotions
- Interactive demo: <https://www.6seconds.org/2022/03/13/plutchik-wheel-emotions/>



# Alternate view: Continuous representation of affect

- Osgood et al. (1957) asked human participants to rate words along dimensions of opposites such as heavy–light, good–bad, strong–weak
- Factor analysis of these judgments revealed that the three most prominent dimensions of meaning:
  - **Valence**/Evaluation/Sentiment (good–bad)
  - **Dominance**/Power/potency (strong–weak)
  - **Activity**/Agency (active–passive)

# Alternate view: Continuous representation of affect



- Emotions can be mapped to these continuous dimensions, rather than being basic discrete categories
- Recall “gender subspace” idea: This seems well-suited to word embeddings? [Sort of works depending on embedding quality Field&Tsvetkov 2019]

# A note on ethics of AI for Emotion Detection

- Plethora of work on using AI for emotion detection and existence of commercial products that claim to be able to do so (e.g. based on facial recognition)
- Limited evidence that this possible
  - Distinction between true internal emotional state and outward expression
  - Lack of consensus on what emotions are among psychologists
- High misuse potential
  - Faulty AI used to make impact decisions in domains like law, education, and employment
- Privacy
  - People generally find “emotion AI” invasive with little benefit for themselves

Barrett, L. F., Adolphs, R., Marsella, S., Martinez, A. M., & Pollak, S. D. (2019). Emotional Expressions Reconsidered: Challenges to Inferring Emotion From Human Facial Movements. *Psychological Science in the Public Interest*, 20(1), 1-68. <https://doi.org/10.1177/1529100619832930>

Pyle, K. Roemmich, and N. Andalibi, “US job-seekers’ organizational justice perceptions of emotion AI-enabled interviews,” *Proceedings of the ACM on Human-Computer Interaction*, vol. 8, 273 no. CSCW2, pp. 1–42, 2024.

# Lexicons: Manual Construction

# What are lexicons?

- A collection of words
- Words with labels
- Some popular lexicons:
  - Linguistic Inquiry and Word Count (LIWC): <https://www.liwc.app/>
  - NRC Emotion Lexicons: <https://saifmohammad.com/WebPages/NRC-Emotion-Lexicon.htm>
  - NRC-VAD Lexicon: <https://saifmohammad.com/WebPages/nrc-vad.html>
  - Connotations frames of power, agency, and sentiment <https://github.com/maartensap/riveter-nlp>

# When are lexicons useful?

- Less ideal use case:
  - Simple classification model (text expresses “anger” if it has a word from an “anger” lexicon)
  - Classifier typically works much better but lexicons are extremely easy to implement (just have to count words) and very interpretable
- More common use cases:
  - Pre-filtering data
    - (e.g. hate speech has low prevalence in randomly sampled social media posts but we can use lexicons of offensive terms to identify what to annotate)
  - Data collection
    - Tweets or news articles that mention particular events
  - Testing robustness/bias, defining meaningful subsets or axes on a scale (think word embedding metrics)

# LIWC

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- Transparent text analysis program that counts words in “psychologically meaningful categories”
- Origins and motivation:
  - Words that people use are reflective of internal state, hidden intentions, psychological state
  - Walter Weintraub (1981, 1989) hand-counted people’s words in texts (political speeches, medical interviews, etc.) and noticed that first-person singular pronouns (e.g., I, me, my) were reliably linked to people’s levels of depression

# LIWC Categories

- 80(+?) categories:
  - Straightforward language dimensions: articles, pronouns
  - More subjective dimensions: emotions, power,
  - Hierarchy of dictionaries:
    - “Anger” dictionary is a subset of “emotion” dictionary

<https://www.liwc.app/static/documents/LIWC-22%20Manual%20-%20Development%20and%20Psychometrics.pdf>



# LIWC Construction

1. Word Collection: “judges brainstorm words for each category” (later versions of LIWC use earlier versions as starting point)
2. Judge Rating phase: 3-4 judges rate “goodness of fit” for each word for each category
3. Base Rate Analysis: Examine word frequency across corpora and remove infrequent words
4. Candidate Word Generation: Examine most frequent words in corpora and determine if they should be added to the Dictionary
5. Psychometric Evaluations: compute internal consistency statistics for each category and manually judge if words “detrimental to the internal consistency” should be omitted (judgements made by the 4 authors)
6. Refinement: Repeat steps 1-5 and check for mistakes
7. Addition of summary variables: add in categories that are summaries of others (e.g. emotional tone)

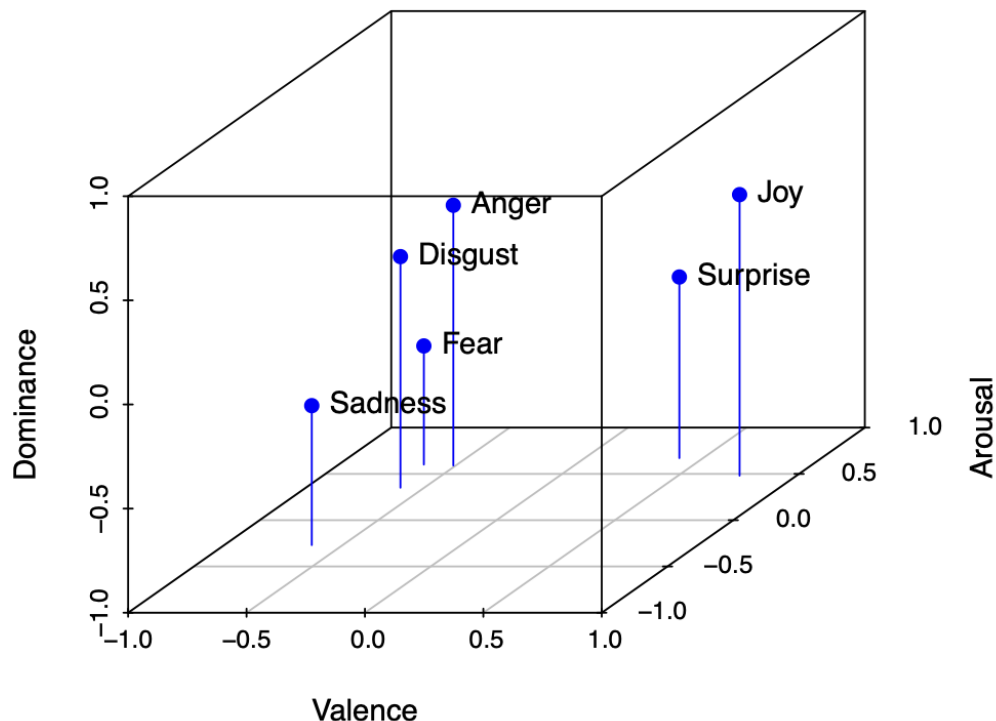
# LIWC Takeaways

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- Really popular resource:
  - Often preferred by social scientists because it was developed by psychologists
  - Commercial easy-to-use software where you can just upload texts and get scores
- Example of data set construction:
  - Relies on domain expertise, judgements of authors and domain experts (not just outsourcing to crowd workers)
  - Iterative process
- Often misused in scenarios it was not designed or evaluated for

# Different Annotation Approach: VAD Lexicons

- LIWC defines discrete categories
- We might want more continuous ratings:
  - Is “annoyed” word associated with “anger”? {0, 1}
  - *How* associated is “annoyed” with “anger”? [0, 1]



# Different Annotation Approach: VAD Lexicons

- Likert Rating scale:

Statement

Academic detailing is a useful form of education that aligns providers' prescribing behavior with evidence-based practice.

Strongly Disagree	Disagree	Neutral	Agree	Strongly Agree
1	2	3	4	5

- Problems:

- Fixed granularity
- Difficult to maintain consistency across annotators
- Difficult for an annotator to be self-consistent
- Scale region bias

# Best-Worst Scaling

Out of these four words (A, B, C, and D):  
Which word is associated with the most/highest valence?  
Which word is associated with the least/lowest valence?

- By answering just these two questions, five out of the six inequalities are known:
  - Example: If A: highest valence and D: lowest valence
    - We know:  $A > B$ ,  $A > C$ ,  $A > D$ ,  $B > D$ ,  $C > D$

# Best-Worst Scaling

$$score(w) = \frac{\#best(w) - \#worst(w)}{\#annotations(w)}$$

- Scores range from -1 to 1
- Empirically shown that three annotations each for  $2N$  4-tuples is sufficient for obtaining reliable scores (where  $N$  is the number of items)

# Score reliability: *split-half reliability* (*SHR*)

- Split all annotations for an item (e.g. 4-tuples) into two halves
- Produce two sets of scores independently from the two halves
- Calculate correlation between the two sets of scores. If the annotations are of good quality, then the correlation between the two halves will be high.
- [Repeat for many, e.g. 100 trials]

# NRC lexicons

1a. **NRC Word-Emotion Association Lexicon** (also called NRC Emotion lexicon or EmoLex). [README](#). Explore the [interactive visualization](#). [Homepage](#) of the Lexicon. Also available in over 40 other languages [here](#). The [sense-level annotations](#) provided by individual annotators for the eight emotions can also be obtained.

1b. **NRC Emotion Intensity Lexicon (aka Affect Intensity Lexicon)**, created using [Best-Worst Scaling](#). The NRC Emotion Intensity Lexicon is a list of English words and their associations with eight basic emotions (anger, anticipation, disgust, fear, joy, sadness, surprise, trust). [Lexicon homepage](#).

2. **NRC Valence, Arousal, Dominance Lexicon**, created using [Best-Worst Scaling](#). The NRC Valence, Arousal, Dominance Lexicon is a list of English words and their valence, arousal, and dominance scores. [Lexicon homepage](#).



# Quiz

## 1. If you are drawing from discrete emotion theory to build an NLP model, what method might you start with?

- A. Word embeddings, where discrete emotions can be represented in high-dimensional affective subspaces
- B. A classification model, where the classifier can output a specific emotion label
- C. LDA Topic modeling, where you can allow each data point to have mixtures of emotions similar to mixtures of topics

## 2. Which of these is NOT a method for constructing annotated lexicons?

- A. Brainstorming starter words for categories and refining them through expert ratings
- B. Likert scale ratings
- C. Best-Worst Scaling
- D. Split-half reliability



## 3. What might be a good use case for LIWC lexicons?

- A. Assessing if restaurant reviewers use linguistic indicators of trauma (e.g. 3<sup>rd</sup> person pronouns, emotion-laden words)
- B. Classifying which tweets about the Black Lives Matter movement express anger
- C. Identifying issue-specific frames in datasets of political speech

# Connotation Frames of Power, Agency, and Sentiment

- Lexicon labels can be discrete or continuous, but they can also be directed
- Connotation frames are a formalism for analyzing subjective roles and relationships implied by a given predicate

"X dismisses Y"

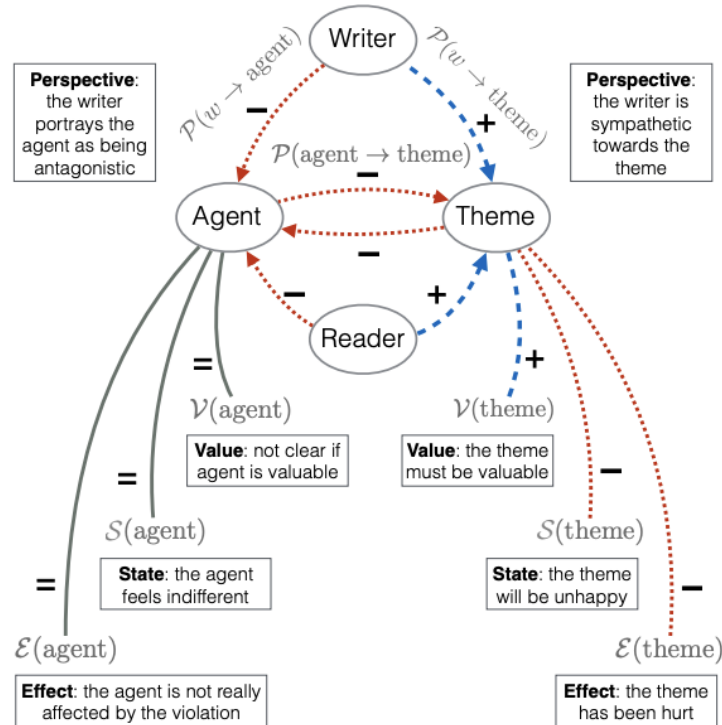
- Writer's Perspective: the writer treats Y more sympathetically but thinks of X as more of an antagonist
- Reader's Perspective: the reader will likely feel sympathetic towards Y and think more poorly of X
- X and Y's Mental State: X may feel indifferent. Y will feel distressed
- X and Y's Perspective, X and Y's Value, Effect on X and Y

Rashkin, Hannah, Sameer Singh, and Yejin Choi. "Connotation Frames: A Data-Driven Investigation." *ACL*. 2016.

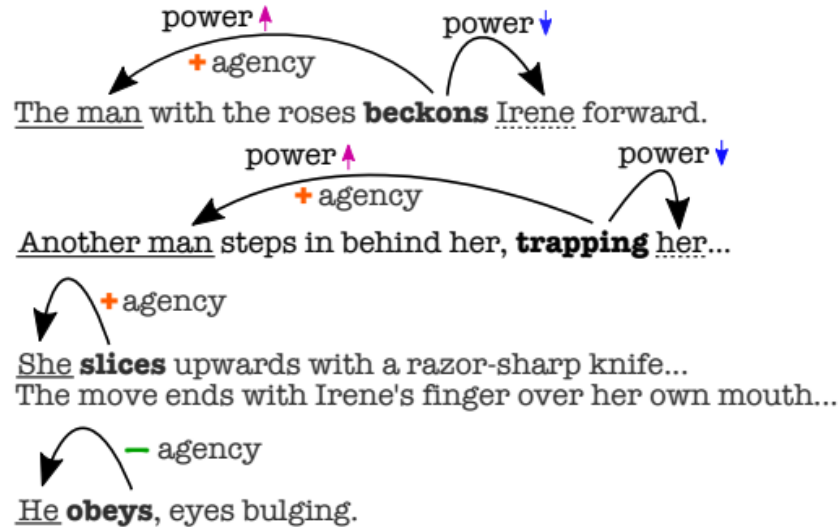
Maarten Sap, Marcella Cindy Prasettio, Ari Holtzman, Hannah Rashkin, and Yejin Choi. 2017. [Connotation Frames of Power and Agency in Modern Films](#). EMNLP. 2017

# Connotation Frames of Power, Agency, and Sentiment

Writer: "Agent violates theme."



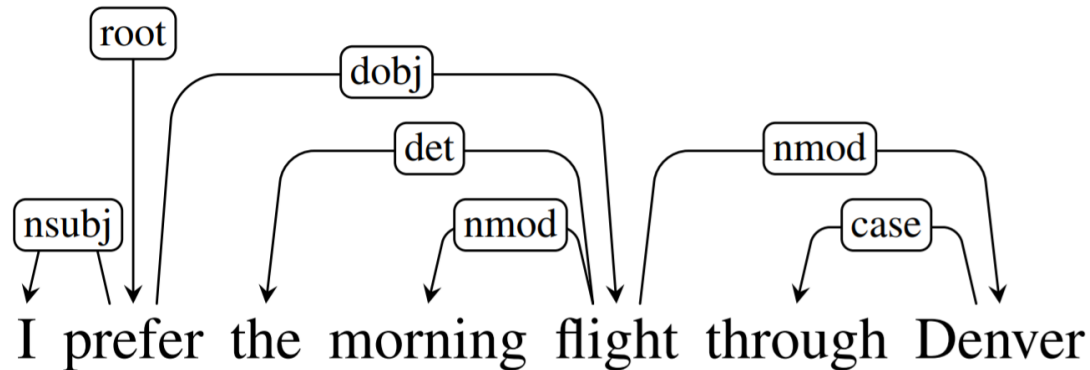
# Connotation Frames of Power, Agency, and Sentiment



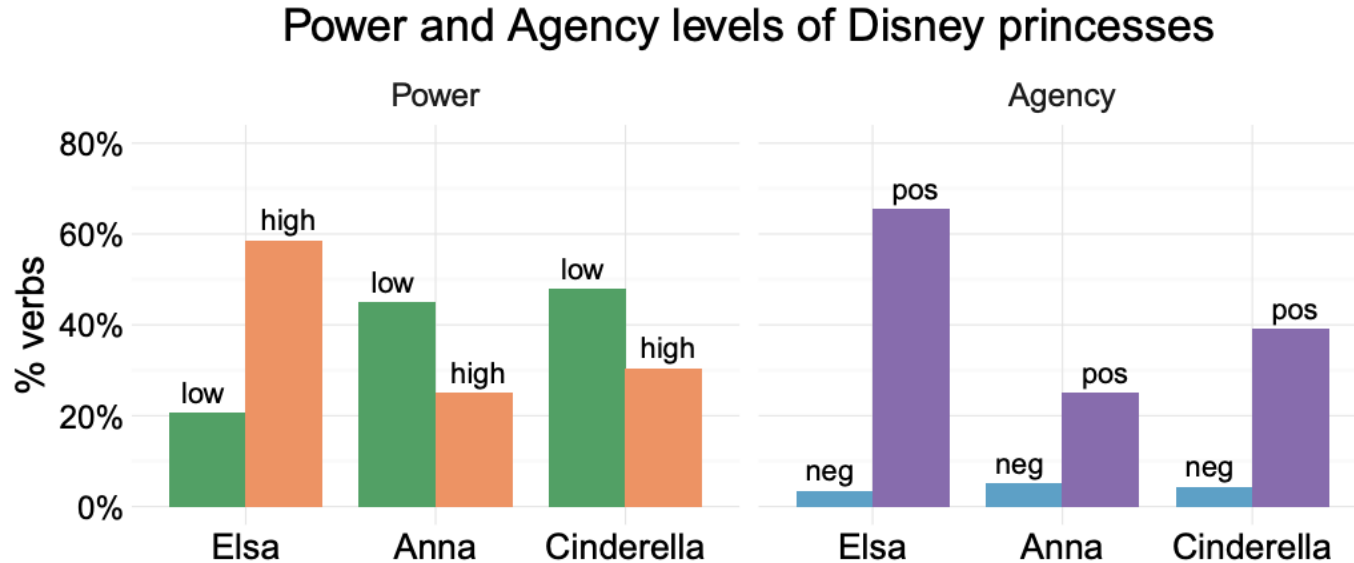
Unlike agency, power is considered to be relative: one entity has power over the other

# Connotation Frames of Power, Agency, and Sentiment

- How do we use annotated connotation frames?
  - We can't just count verbs – we need to resolve agent/theme or subject/object
- Dependency parsing (alternative: semantic role labeling):



# Connotation frames: Movie Analysis



# Lexicons: Automated Construction

# Inducing Domain-specific lexicons

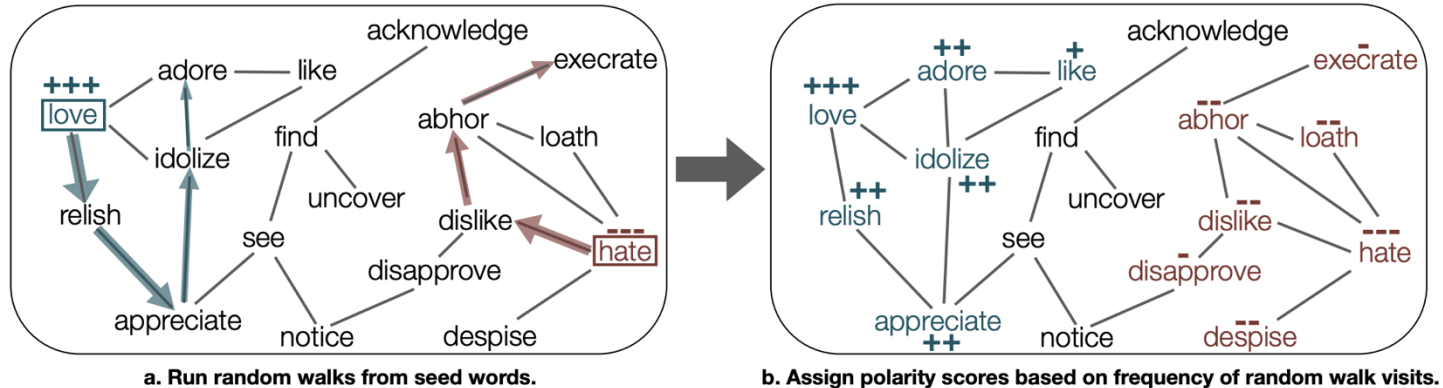
- A word's sentiment (or connotation or emotion) depends on the domain in which it is used
  - Words can change meaning over time
  - Connotations can be domain-specific: NRC lexicons associate "police" with "trust"
    - Not the association you would expect in a social movement about police brutality
- What can we do about this?
  - Annotate a new lexicon for every domain of interest? → Time consuming and expensive



# SentProp: Algorithm for Domain-specific sentiment lexicons

- Starting point: small seed set of negative and positive words (e.g. ~10 each)
- Construct word embeddings (they use matrix-decomposition approach)
- Construct a graph representation
  - Words are nodes
  - Edges are between each node's k-nearest neighbors (based on embedding similarity)
  - Run a random walk (with transition matrix defined by edges)
  - Polarity scores are based on random walk visits

# SentProp: Algorithm for Domain-specific sentiment lexicons



**Figure 3:** Visual summary of the SENTPROP algorithm.

- Evaluation: recreating existing lexicons

# Example differing domain-induced lexicons: two subreddits



# Alternative approaches to lexicon induction

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- Word co-occurrence PMI scores (Turney and Littman, 2003)
- Variants of the propagation approach or embedding construction (Velikovich et al. 2010)
- DENSIFIER (Rothe et al. 2016): condenses word embeddings into a single dimension

# Recap

- Emotions:
  - Different models of emotions in psychology
- Lexicons:
  - Commonly used lexicons
    - LIWC, NRC lexicons, connotation frames
  - When lexicons are useful and when they are not
  - Different ways of constructing them
    - Manual vs. automated, categorical vs. continuous, directed (connotation frames) vs. not
- Data annotating:
  - Likert scale, Best-worst scaling

# Logistics

- HW 1 is due Monday

# References

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- Giovanna Colombetti (2009) From affect programs to dynamical discrete emotions, *Philosophical Psychology*, 22:4, 407, DOI: [10.1080/09515080903153600](https://doi.org/10.1080/09515080903153600)
- [Obtaining Reliable Human Ratings of Valence, Arousal, and Dominance for 20,000 English Words](#). Saif M. Mohammad. ACL 2018.
- Tausczik, Yla R., and James W. Pennebaker. "The psychological meaning of words: LIWC and computerized text analysis methods." *Journal of language and social psychology* 29.1 (2010): 24-54.
  - <https://www.liwc.app/static/documents/LIWC-22%20Manual%20-%20Development%20and%20Psychometrics.pdf>