TEXT VISUALIZATION

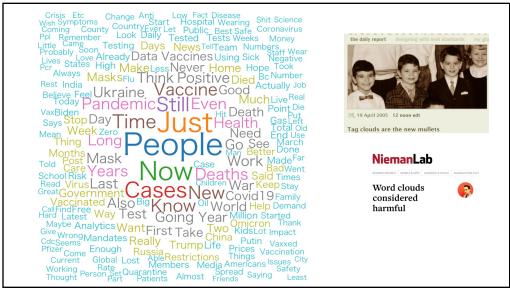
CS 448B | Fall 2024

MANEESH AGRAWALA

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READING RESPONSE: QUESTIONS/THOUGHTS

While a method like ReVision (created in a research environment) would likely not be widely used until its accuracy improves significantly, I wonder if there are already AI chart generation or redesign tools being used that might not have perfect accuracy (and that are not subject to as much scrutiny as university research tools). Who is responsible for mistakes or misinformation in published charts? Is it the data collector, the creator of the visualization, or the editor?

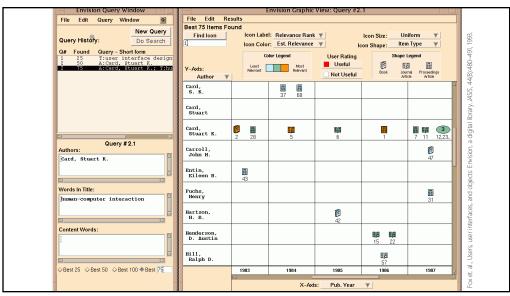


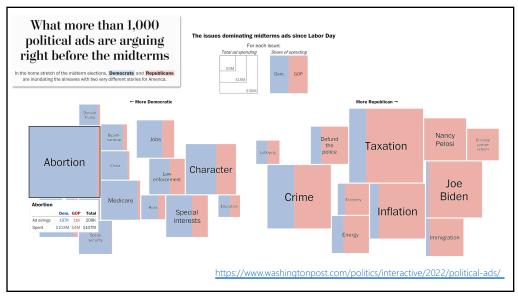


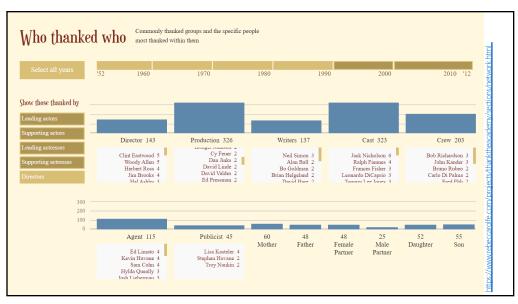
TASKS

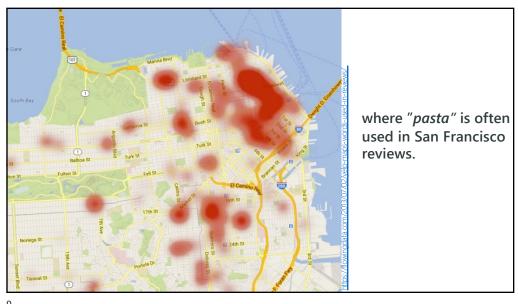
- What documents contain text about X?
- Which documents are of interest to me?
- Are there documents similar to this one?
- How are different words used in the collection?
- What are the main themes?
- How are themes distributed?
- What unifies/connects documents in a collection?

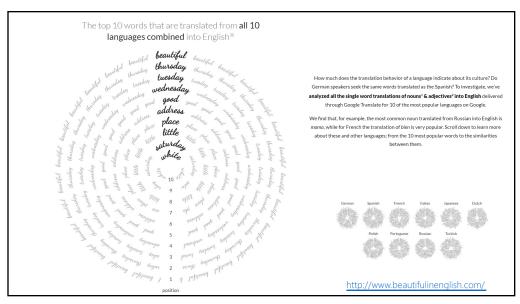
5











Learning Objectives

TODAY

- 1. Considering text as data
- 2. Identifying descriptive words/keyphrases
- 3. Visualizing text in context
- 4. Searching across and comparing documents

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TEXT AS DATA

1/

TEXT AS NOMINAL DATA

High dimensional (many words, e.g. 10,000+ common words)

More than equality tests

- Correlations: Hong Kong, San Francisco, Bay Area
- Order: April, February, January, June, March, May
- Membership: Tennis, Running, Swimming, Hiking, Piano
- Hierarchy, antonyms & synonyms, entities, ...

Words have meanings and relations

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TEXT PROCESSING PIPELINE

Tokenization

Segment text into terms
Remove stop words? a, an, the, of, to, be
Numbers and symbols? #cardinal, @Stanford, OMG!!!!!!!!
Entities extraction? Palo Alto, O'Connor, U.S.A.

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Stemming

Group together different forms of a word
Porter stemmer? visualization(s), visualize(s), visually → visual
Lemmatization? goes, went, gone → go

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Ordered list of terms

IDENTIFYING DESCRIPTIVE WORDS/KEYPHRASES

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BAG OF WORDS MODEL

Ignore ordering relationships within the text

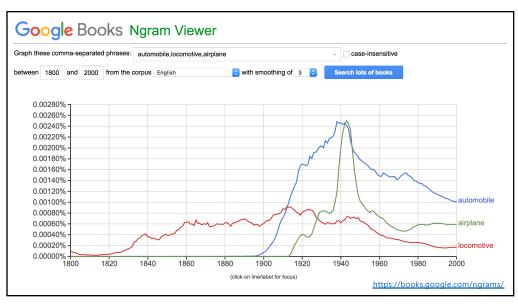
A document ≈ vector of term weights

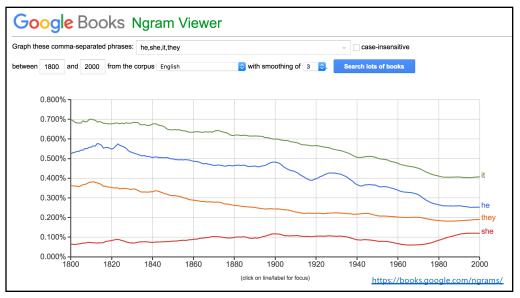
- Each term corresponds to a dimension (10,000+)
- Each value represents the relevance For example, simple term counts

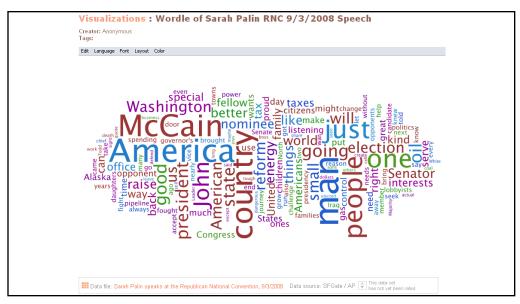
Aggregate into a document x term matrix

• Document vector space model

DOCUMENT X TERM MATRIX Each document is a vector of term weights Simplest weighting is to just count occurrences Antony and Cleopatra Julius Caesar The Tempest Hamlet Othello Macbeth Antony **Brutus** Caesar Calpurnia Cleopatra mercy worser







WORD/TAG CLOUDS

Strengths

- Compact lots of words fit
- Draws eye to most frequent/biggest words
- Can help with gisting and initial query formation

Weaknesses

- Sub-optimal visual encoding (size not pos. encodes freq.)
- Inaccurate size encoding (long words are bigger)
- May not facilitate comparison (unstable layout)
- Term frequency may not be meaningful
- Does not show the structure of the text

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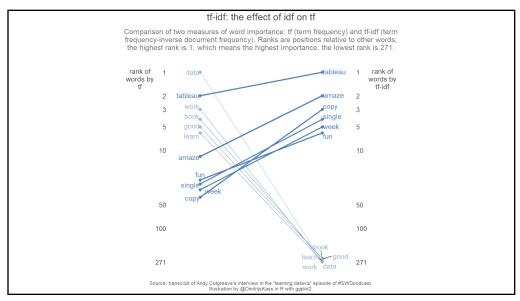
WORD WEIGHTING

Term Frequency $tf_{td} = count(t)$ in document

TF.IDF: Term Frequency by Inverse Document Freq

 $tf.idf_{td} = log(1 + tf_{td}) \times log(N/df_t)$

 $df_t = \#$ documents containing term t N = # of documents



LIMITATIONS OF FREQUENCY STATISTICS

Typically focus on unigrams (single terms)

Often favors frequent (TF) or rare (IDF) terms
Not clear that these provide best description of documents

"Bag of words" ignores additional info

Grammar / part-of-speech Position within document Recognizable entities





TIPS FOR DESCRIPTIVE KEYPHRASES

Understand the limitations of your language model

Bag of words:

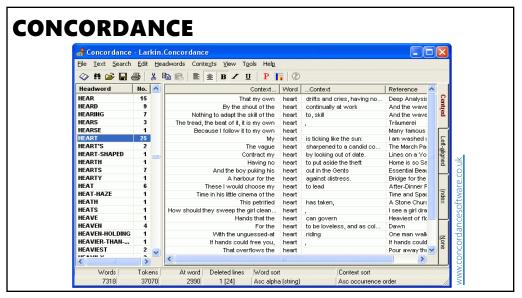
Easy to compute
Single words
Loss of word ordering

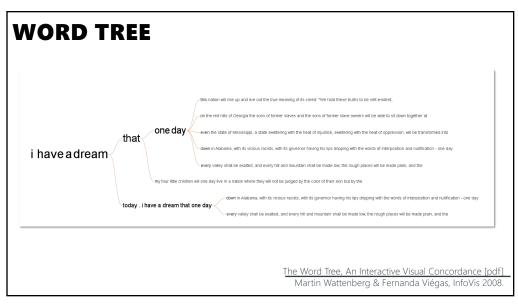
Select appropriate model and visualization

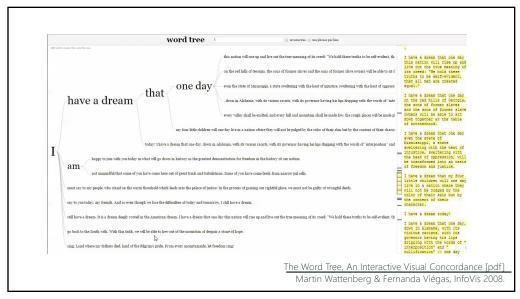
Generate longer, more meaningful phrases Adjective-noun word pairs for reviews

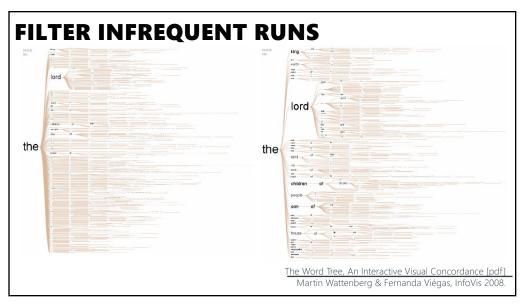
34

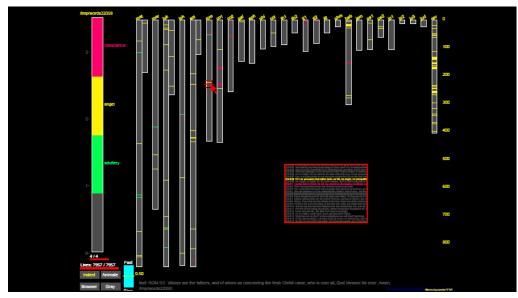
VISUALIZING WORDS/KEYPHRASES CONTEXT

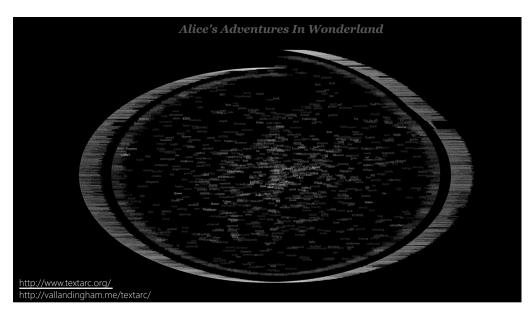


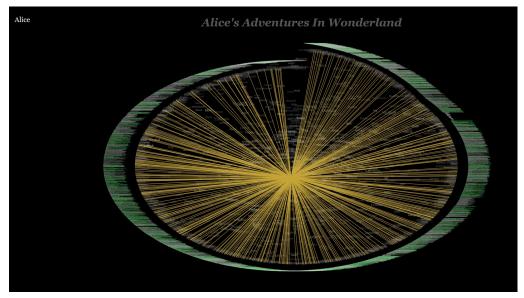




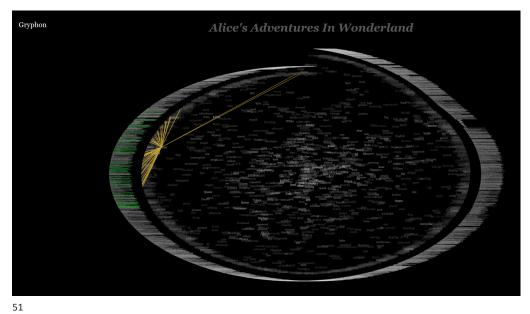














GLIMPSES OF STRUCTURE

Concordances and TextArc show local, repeated structure But what about other types of patterns?

For example

Lexical: <A> at

Syntactic: <Noun> <Verb> <Object>

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PHRASE NETS

Look for specific linking patterns in the text:

'A and B', 'A at B', 'A of B', etc. Could be output of regexp or parser

Visualize extracted patterns in a node-link view

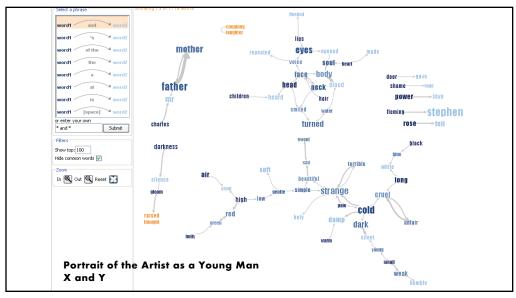
Occurrences → Node size

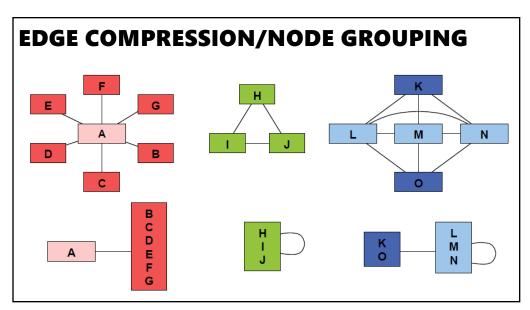
Pattern position → Edge direction

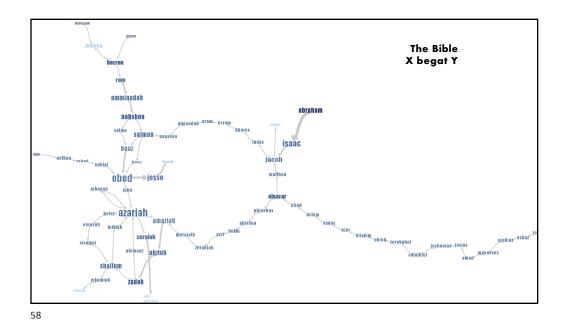
Darker color → higher ratio of out-edges to in-edges

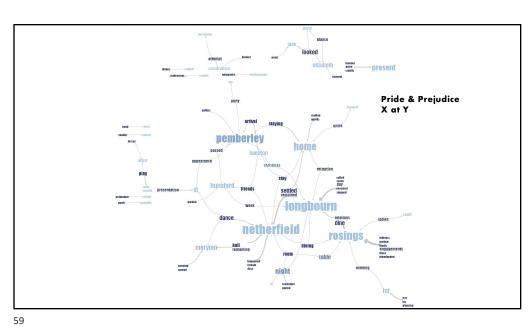
Mapping Text with Phrase Nets

Frank Van Ham, Martin Wattenberg & Fernanda Viégas, InfoVis 2009.

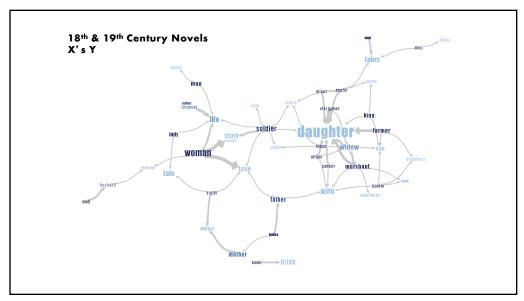


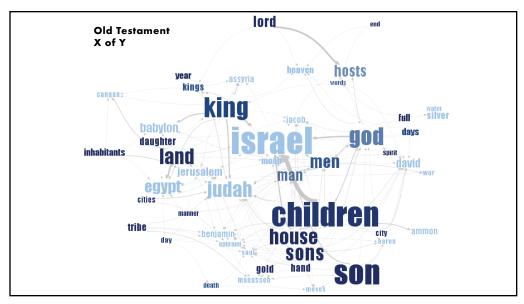


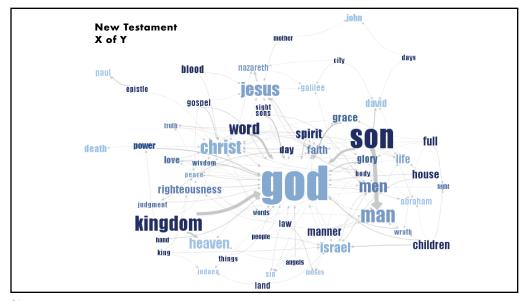




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FINAL PROJECT

Design Reviews Dec 2 and Dec 4 (signups this week)

Data analysis/explainer

Analyze dataset in depth & make a visual explainer

Deliverables

An article with multiple different interactive visualizations Short video (2 min) demoing and explaining the project

Schedule

Design Review and Feedback: 10^{th} week of quarter, 12/2 and 12/4 Final code and video: Sun 12/8 8pm

Grading

Groups of up to 3 people, graded individually Clearly report responsibilities of each member

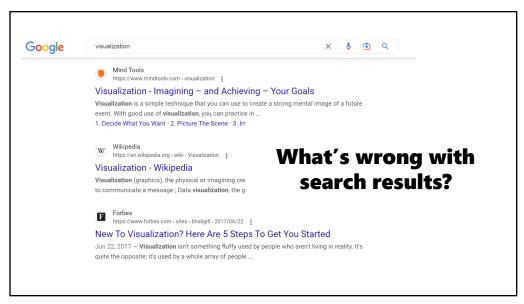
66

DESIGN REVIEW SIGNUPS

Sign up for 8-10 min slot with teaching team

Will offer slots on Mon 12/2 & Wed 12/4 during class period and perhaps others Stay tuned for a canvas announcement about signup sheet





PROBLEMS WITH SEARCH RESULTS

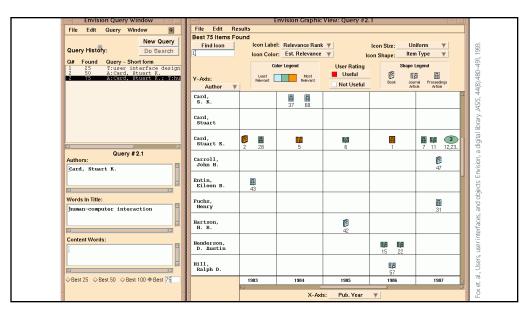
Query responses don't tell you:

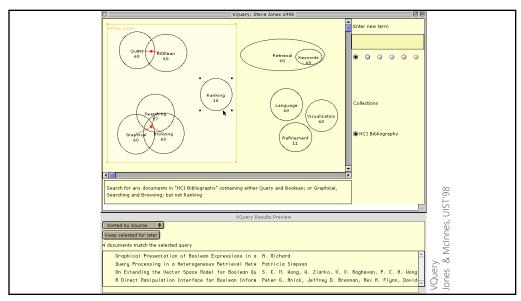
- How strong the match
- How frequent each term
- · How term is distributed
- Overlap between terms
- Length of document

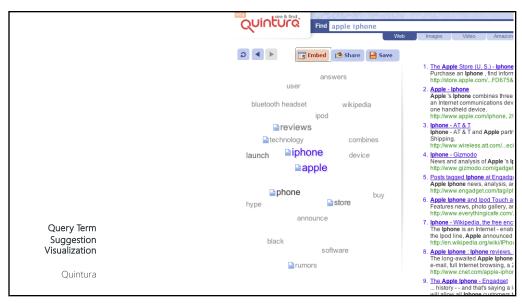
Ranking is opaque Inability to compare results

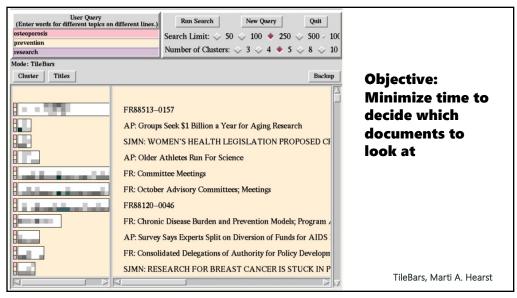
Marti A. Hearst

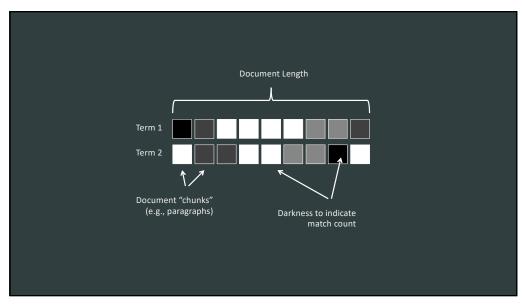
70

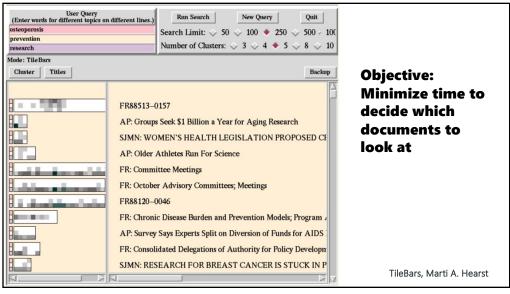


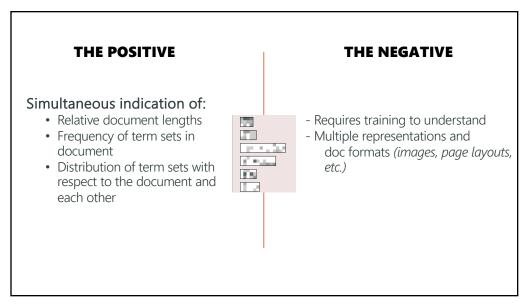












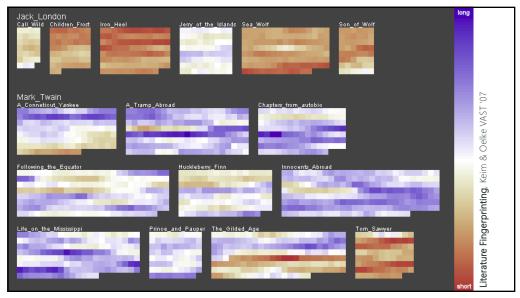
COMPARING DOCUMENTS

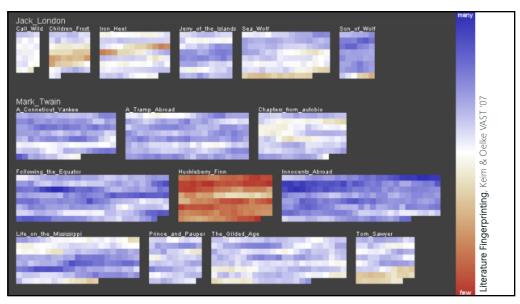
81

LITERARY ANALYSIS

Features for comparison

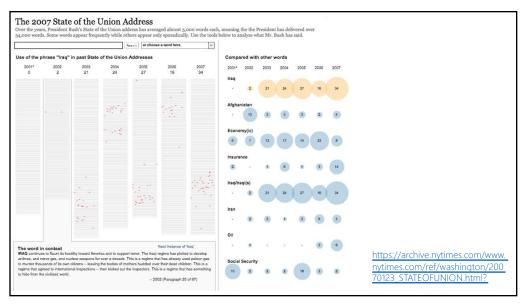
- Word length
- Syllables per word
- Average sentence length
- Percentage by parts of speech (nouns, verbs, etc.)
- Frequencies of specific words
- Hapax Legomena (words that appear once)

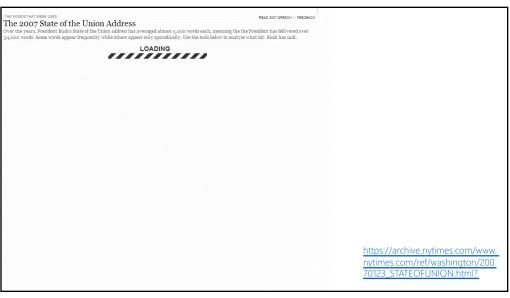


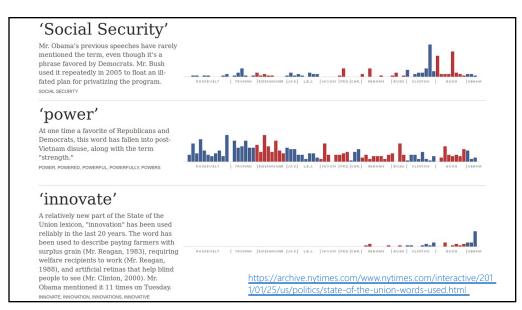


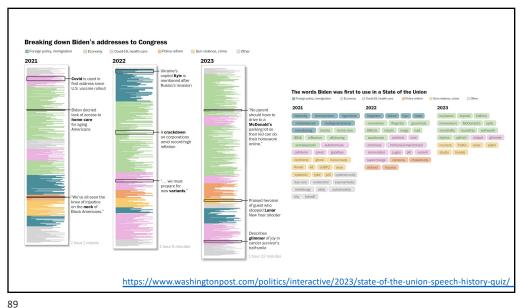
表記事時期發展的表面發展的表現的 (A) 1919年後的 (B) 1919年 (B) 19 BESELDE ER BESELDE BES · 医克里克斯氏征 1985年 1985年 1988年 DIO DE LA CONTRACTOR DE an litering belande in and the state of the Literature Fingerprinting, Keim & Character fürziebilit igetise gebilekroberbliebiletenbilebiletiebilebilebilebilebilebilebenbebeitenenbeiteb 表示是以自由的自己的对象的。 ngoughmanngongungungungung ömenmengung forpresporet fün betrappet biffe frente fie fie bif ib befere Badde ebaserbaterent gerrilkibberides gittelterter nobekt geunde beut beit fielt git iter Be filt it: a cittiffichides geit eant ab l'eac i a l'eactheallisteadh

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VISUALIZING CONVERSATIONS

MANY DIMESIONS TO CONSIDER

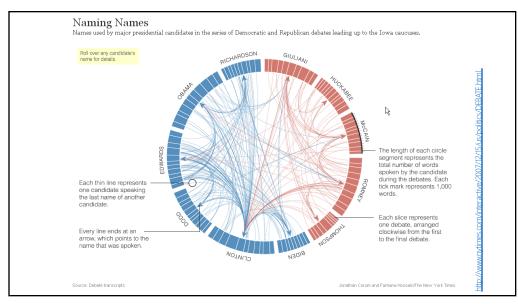
Dimensions

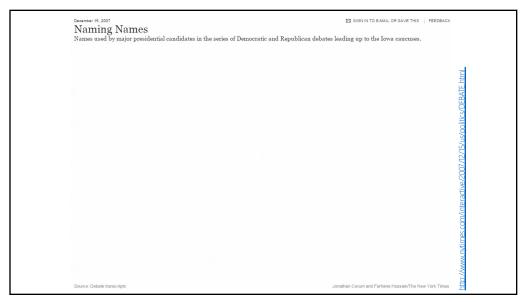
- Who (senders, receivers)
- What (the content of communication)
- When (temporal patterns)

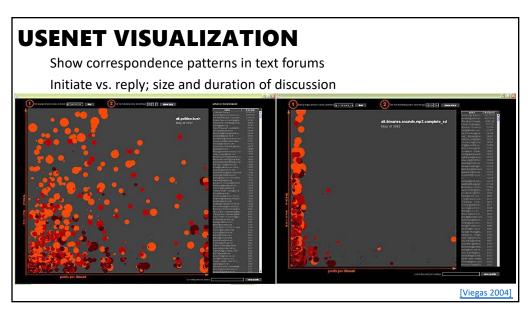
Cross-products

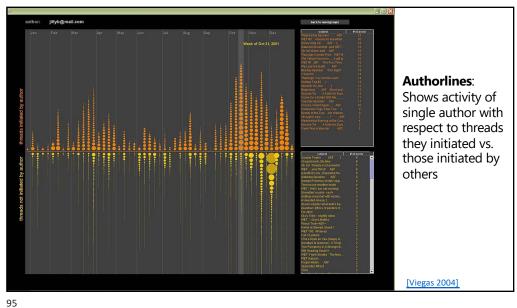
- What x When → Topic "Zeitgeist"
- Who x Who → Social network
- Who x Who x What x When → Information flow

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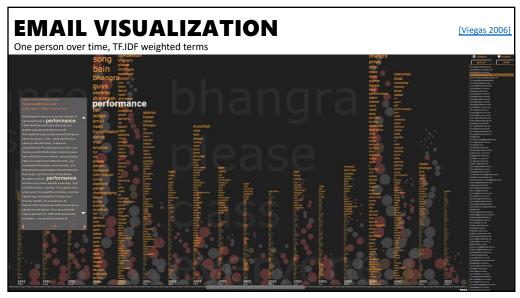












SUMMARY

Text has many levels of possible visualizations

• Word, document, collection

Identifying descriptive words/keyphrases is critical

- e.g., TF, TF.IDF, regexp, ...
- Domain dependent

Can go beyond standard charts and graphs but requires vigilance in design decisions (don't be seduced by possibilities)