VISUALIZING TEXT

Petra Isenberg



RECAP

STRUCTURED DATA



0.103 0.176 0.387 0.300 0.379
0.333 0.384 0.564 0.587 0.857
0.421 0.309 0.654 0.729 0.228
0.266 0.750 1.056 0.936 0.911
0.225 0.326 0.643 0.337 0.721
0.187 0.586 0.529 0.340 0.829
0.153 0.485 0.560 0.428 0.628

UNSTRUCTURED DATA





(TODAY)

VISUALIZING TEXT

amet, consectetus adipirique elit, sed do eiusmod tempor incididunt ut labore et dolore magna aliqua. Ut enini ad minim veniam, quis nostrud exercitation ullamco laboris nici paliquipos ca commodo consequat. Duis aute irure dolor in reprehenderit in pariatur esse cillum dolore eu fugiat nulla pariatur. Excepteur sint orgaers repjdent, project in culpa qui officia deserunt mollit anim il est laboram. Lorem psum color sit amet, consectetur adipisicing elit, sed de cius politorin proprieta di la la la cius politorin di la la la cius politorin di la cius politorin di la la cius politorin di la c ut aliquip ex ea compodo consequat. Duis aute irure dolor in reprehenderit in voluptate velit esse ellum dolore eu fugiat nulla pariatur. Excepteur sint occaecat cupidatat non proident, sunt in culpa qui officia deserunt mollit anim id est laborum. Lorem ipsum dolor sit amet, consectetur adipisicing elit, sed do eiusmod tempor incididunt ut labore et dolore magna aliqua. Ut enim ad minim veniam, quis nostrud exercitation ullamco laboris nisi ut aliquip ex ea commodo

consequat. Duis aute irure dolor in reprehenderit in voluptate velit esse cillum

nostrud exercitation ullamco laboris nisi ut aliquip ex ea commodo consequat.

Duis aute irure dolor in reprehenderit in voluptate velit esse cillum dolore eu

fugiat nulla pariatur. Excepteur sint occaecat cupidatat non proident, sunt in

culpa qui officia deserunt mollit anim id est laborum. Lorem ipsum dolor sit

TEXT?

WHY

- To assist information retrieval
- To enable linguistic analysis
- To augment analytics on mixed data



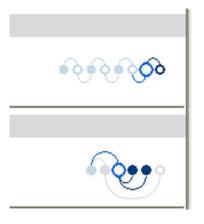
Visual Theosento

Service Production of the Interaction

Service Production of the Interaction

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Themescape

Visual Thesaurus

Thread Arcs

WHY

UNDERSTANDING: GET THE "GIST" OF A DOCUMENT

GROUPING: CLUSTER FOR OVERVIEW OR CLASSIFICATION

COMPARE: COMPARE DOCUMENT COLLECTIONS, OR INSPECT EVOLUTION OF COLLECTION OVER TIME

CORRELATE: COMPARE PATTERNS IN TEXT TO THOSE IN OTHER DATA, E.G., CORRELATE WITH SOCIAL NETWORK

WHAT IS TEXT

DOCUMENTS

ARTICLES, BOOKS AND NOVELS COMPUTER PROGRAMS E-MAILS, WEB PAGES, BLOGS TAGS, COMMENTS

COLLECTION OF DOCUMENTS

MESSAGES (E-MAIL, BLOGS, TAGS, COMMENTS)
SOCIAL NETWORKS (PERSONAL PROFILES)
ACADEMIC COLLABORATIONS (PUBLICATIONS)
EVEN WHOLE LIBRARIES, WEBSITES, SOCIAL NETWORKS

DIFFICULT DATA

- Too much data what to use?
 - Millions of blog posts,
 - Hundreds of thousands of news stories,
 - 183 billion emails,
 - ... per day
- Data is noisy:
 - 70-72% of email is spam
 - Text contains section headings, figure captions, and direct quotes
 - **—**

ONCE YOU HAVE THE DATA...

- Most meaning comes from our minds and common understanding.
- "How much is that doggy in the window?"
 - how much: social system of barter and trade (not the size of the dog)
 - "doggy" implies childlike, plaintive, probably cannot do the purchasing on their own
 - "in the window" implies behind a store window, not really inside a window, requires notion of window shopping

(Hearst, 2006)

LANGUAGE IS AMBIGUOUS

- Words and phrases can have many meanings, determined by context and world knowledge
- Interesting language is often figurative:
 - America is a melting pot (metaphor)
 - Busy as a bee (simile)
 - Opportunity knocked on the door (personification)
 - You could have knocked me over with a feather (hyperbole)

LANGUAGE IS AMBIGUOUS

"I can't tell you how much I enjoyed meeting your husband."

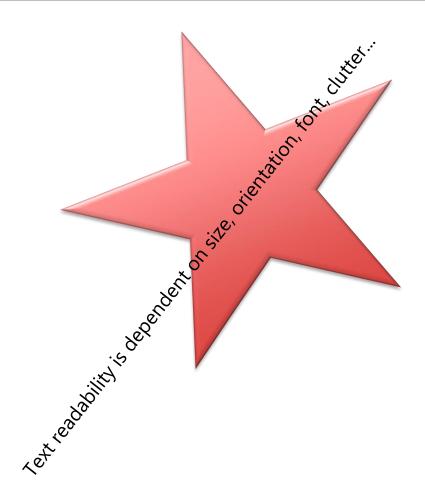
(William Empson, Seven Types of Ambiguity, 1947)

"Brave men run in my family."

(Bob Hope as "Painless" Peter Potter in The Paleface, 1948)

Supporters of Martin, who has been jailed without trial for more than two years, are calling on Prime Minister Stephen Harper to ask Mexican president Felipe Calderon to release Martin text is not preattentive under a section of the Mexican constitution that allows the government to expel undesirables from the country. Martin's supporters believe she has no chance of a fair trial in Mexico. Neither does Waage.

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- Text readability is dependent on size, orientation, font, clutter...
- More likely to need large amounts of text in language visualization

VISUALIZING LANGUAGE IS ALSO EASY!

- SO much data available for analysis
- (Mostly) readily computer readable
- Simple techniques can give instant summaries

OUTLINE

- TEXT AS DATA
- VISUALIZING DOCUMENT CONTENT
- EVOLVING DOCUMENTS
- DOCUMENT COLLECTIONS

TEXT AS DATA

Words are the basic unit of data.

WORD-LEVEL ATTRIBUTES

- WORD LENGTH
- PART OF SPEECH (NOUN, VERB, ADJECTIVE, ETC.)
- FORMAT (/TALIC, UNDERLINE, ETC.)
- LANGUAGE (ENGLISH? LATIN? JAPANESE?)
- FREQUENCY / DIFFICULTY (IS IT COMMON?)
- SENTIMENT (POSITIVE OR NEGATIVE CONNOTATION)
- SYNONYMS / ANTONYMS / ETYMOLOGY (OTHER MEANINGS? ROOTS?)
- ENTITIES (e.g. "Calgary", "Obama", "Telus")
- ... AND MANY MORE

AGGREGATION

REPETITION PLAGARISM SHARED ENTITIES AUTHOR STYLE

COLLECTION

- DOCUMENT
 - SECTION
 - PAGE
- PARAGRAPH
- SENTENCE

WORD

TENSE SENTIMENT SENTENCE LENGTH READING I FVFI

LINGUISTIC METHODS

- Word Counting
- Word Scoring
- Stemming
- Stop Word Removal
- Part of Speech Tagging
- Parsing
- Word Sense Disambiguation
- Named Entity Recognition
- Semantic Categorization
- Sentiment Analysis
- Topic Modeling (some caveats)

WHAT ABOUT THESE WORDS?

automate automates automatic automation



a, an, the, to, ...

- " New York
- " Ban Ki-moon
- " Manchester United

STEMMING

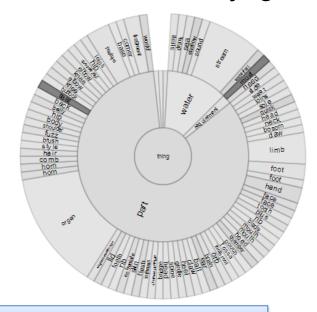
- Reduce words to their 'stems' by removing endings (morphology)
 - running -> run
 - runs -> run
- A good way to increase signal and reduce fracturing of the corpus if there aren't many words.
- Note: Keep the original words somewhere! Also keep the case if you choose to lowercase the word; you never know when you'll need this data

STOP WORD REMOVAL

- Common words such as "and", "the", "I" are removed from view to highlight content words
- Domain specific stop words, e.g. in legal domain:
 - Court, attorney, honour, plaintiff, etc.
- Caution! These words have been shown to be useful for stylistic analysis! When working with text corpora, KEEP EVERYTHING.

NAMED ENTITY RECOGNITION

- What are the people, places in the text?
- Use NLTK it's very good at this.



food tomy displeasure
Hercules
Adam Claudio
prince And How head
counsel Don Pedro

Signior Benedick
Messenger maid
beggar BEATRICE Well
Signior Leonato

Much Ado About Nothing

0 11.67

http://vialab.science.uoit.ca/docuburst

TEXT PROCESSING

TOKENIZATION: SEGMENT TEXT INTO TERMS

ENTITIES? "SAN FRANCISCO", "O'CONNOR", "U.S.A."

REMOVE STOP WORDS? "A", "AN", "THE", "TO", "BE"

N-GRAMS? CAN TAKE WORDS IN 2-WORD GROUPS (BI-GRAMS), 3-WORD (TRI-GRAMS), ETC.

STEMMING: GROUP TOGETHER DIFFERENT FORMS

ROOTS: VISUALIZATION(S), VISUALIZE(S), VISUALLY → VISUAL

LEMMATIZATION: GOES, WENT, GONE → GO

FOR VISUALIZATION, SOMETIMES NEED TO REVERSE STEMMING FOR LABELS

SIMPLE SOLUTION: MAP FROM STEM TO THE MOST FREQUENT WORD

RESULT: ORDERED STREAM OF TERMS

TEXT PROCESSING

"The quick brown fox jumps over the lazy dog."

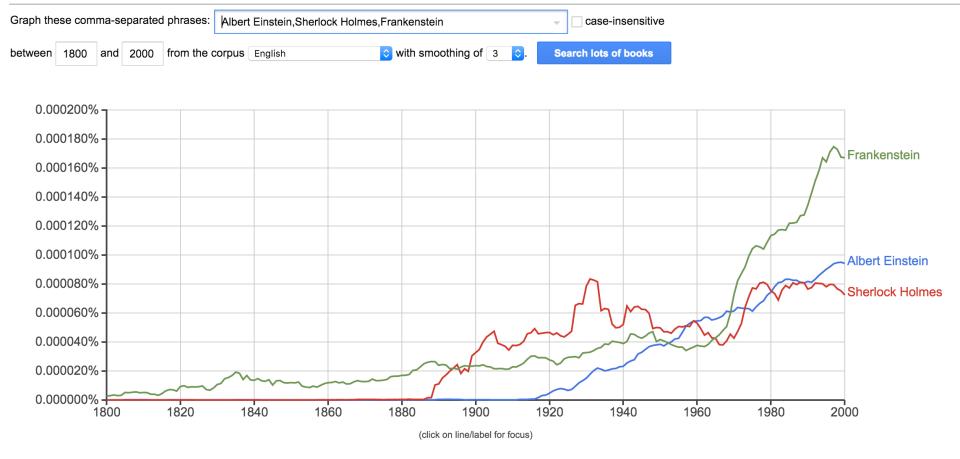
TOKENIZE (N=1)
[The], [quick], [brown], [fox], [jumps], [over], [the], [lazy], [dog].

TOKENIZE (N=1), REMOVE STOPWORDS, STEM [quick], [brown], [fox], [jump], [over], [lazy], [dog]

TOKENIZE (N=2) [the quick], [quick brown], [brown fox], [fox jumps], [jumps over], [over the]...

TOKENIZE (N=5) [the quick brown fox jumps over], [brown fox jumps over].

Google Books Ngram Viewer



NLTK (NATURAL LANGUAGE TOOLKIT)

Tokenize and tag some text:

```
>>> import nltk
>>> sentence = """At eight o'clock on Thursday morning
... Arthur didn't feel very good."""
>>> tokens = nltk.word_tokenize(sentence)
>>> tokens
['At', 'eight', "o'clock", 'on', 'Thursday', 'morning',
'Arthur', 'did', "n't", 'feel', 'very', 'good', '.']
>>> tagged = nltk.pos_tag(tokens)
>>> tagged[0:6]
[('At', 'IN'), ('eight', 'CD'), ("o'clock", 'JJ'), ('on', 'IN'),
('Thursday', 'NNP'), ('morning', 'NN')]
```

Identify named entities:

DOCUMENT CONTENT

BUT FIRST SOME SKETCHING

SKETCHING: VISUALIZE

IMAGINE YOU HAVE 20 YEARS OF SCIENTIFIC PAPER ABSTRACTS:

YEAR

CONFERENCE

AUTHORS

TITI F

PAPER TYPE

KEYWORDS

REFERENCES

ABSTRACT TEXT

TASK:

1) VISUALIZE THE MOST IMPORTANT CONTENT FROM A SINGLE PAPER.

2) VISUALIZE HOW SIMILAR PAPERS FROM EACH CONFERENCE ARE TO PAPERS FROM OTHER CONFERENCES.

GROUPS OF 3

(~10 MINUTES)

EXAMPLE

Tools & Strategies for Social Data Analysis

by

Wesley Jay Willett

A dissertation submitted in partial satisfaction of the requirements for the degree of Doctor of Philosophy

in

Computer Science

in the

Graduate Division

of the

University of California, Berkeley

THESIS WESLEY WILLETT

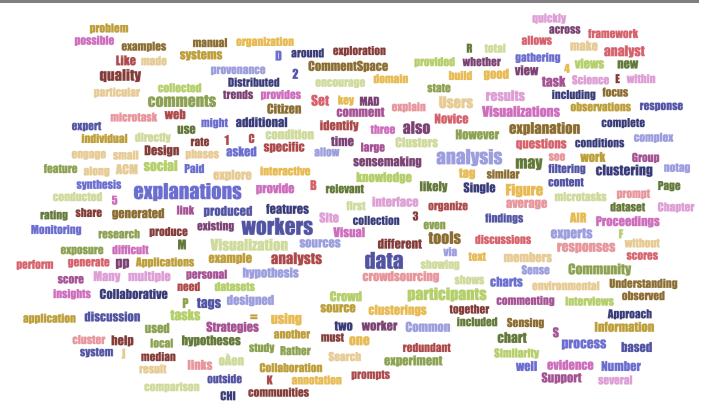
TAG CLOUDS

WORD COUNT

additional air analysis analysts annotation applications approach asked author average based build chart citizen Clustering collaborative collection **COMMENTS** commentspace Community complete Condition contributions ${\sf crowd \ crowdsourcing} \ data {\sf datasets \ design \ different \ discussion \ evidence \ example}$ experiment experts explanations explore features figure filtering generated group help hypotheses hypothesis identify including indicating information interactive interface knowledge links members microtasks multiple novice number oaen observations organize participants phases pp proceedings process produced prompt provide quality questions rate redundant requires responses results score sense share showing similar site SOCial SOURCE specific state strategies study support systems tags tasks tools understanding used users views visualization web work WORKERS

TAG CLOUDS

WORD COUNT



WHAT'S PROBLEMS DO YOU SEE WITH TAG CLOUDS?

additional air analysis analysts annotation applications approach asked author average based build chart citizen clustering collaborative collection comments comments comments community complete condition contributions crowd crowdsourcing data datasets design different discussion evidence example experiment experts explanations explore features figure filtering generated group help hypotheses hypothesis identify including indicating information interactive interface knowledge links members microtasks multiple novice number oaen observations organize participants phases pp proceedings process produced prompt provide quality questions rate redundant requires responses results score sense share showing similar site social source specific state strategies study support systems tags tasks tools understanding used users views visualization web work workers



TAG CLOUDS

STRENGTHS

CAN HELP WITH GISTING AND INITIAL QUERY FORMATION.

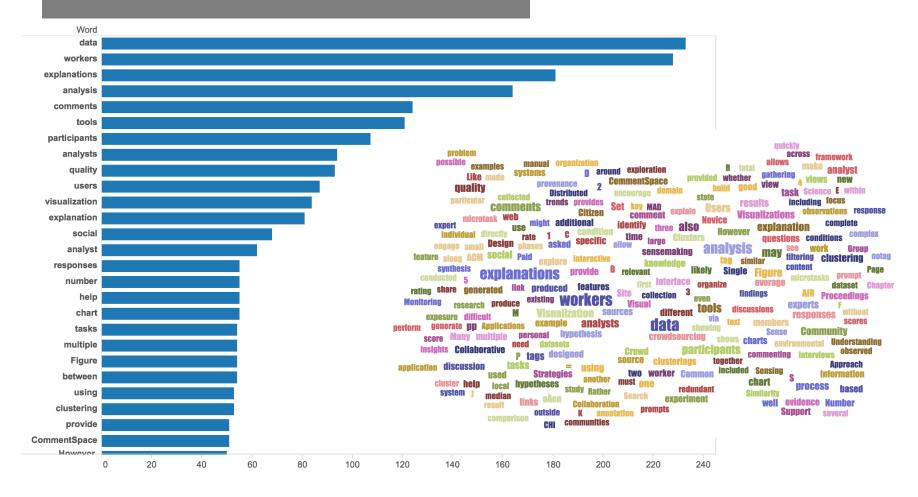
WEAKNESSES

SUB-OPTIMAL VISUAL ENCODING (SIZE VS. POSITION)
INACCURATE SIZE ENCODING (LONG WORDS ARE BIGGER)
MAY NOT FACILITATE COMPARISON (UNSTABLE LAYOUT)

 ORDER USUALLY MEANINGLESS (USUALLY ALPHABETICAL OR RANDOM)

TERM FREQUENCY MAY NOT BE MEANINGFUL DOES NOT SHOW THE STRUCTURE OF THE TEXT

WORD COUNTS



WORDCOUNT

WORDCOUNT



CURRENT WORD

FIND WORD:

BY RANK:

REQUESTED WORD: THE

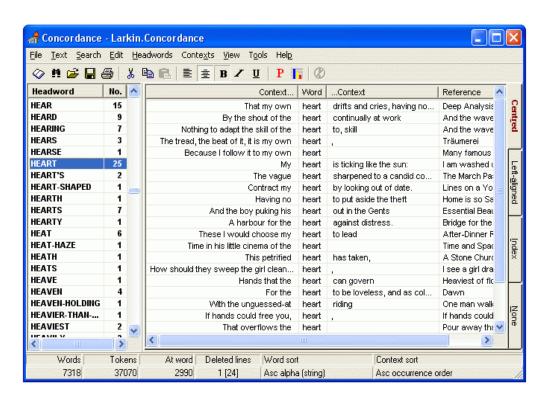
RANK: 1



http://wordcount.org

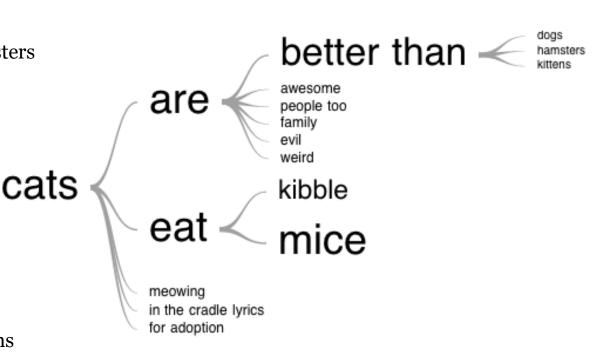
CONCORDANCE

WHAT IS THE COMMON LOCAL CONTEXT OF A TERM?



WORD TREES

- cats are better than dogs
- cats eat kibble
- cats are better than hamsters
- cats are awesome
- cats are people too
- cats eat mice
- cats meowing
- cats in the cradle
- cats eat mice
- cats in the cradle lyrics
- cats eat kibble
- cats for adoption
- cats are family
- cats eat mice
- cats are better than kittens
- cats are evil
- cats are weird
- cats eat mice



WATTENBERG & VIÉGAS 2008

31



hath bestowed upon us, that we should be called the sons of god: therefore the world knoweth us not, because it knew him

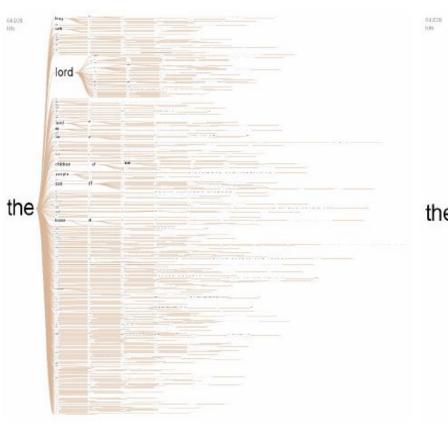
brotherhood

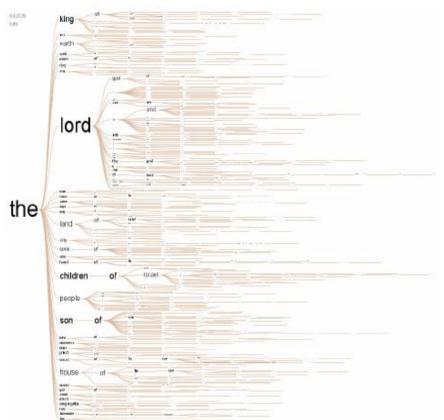
brethren

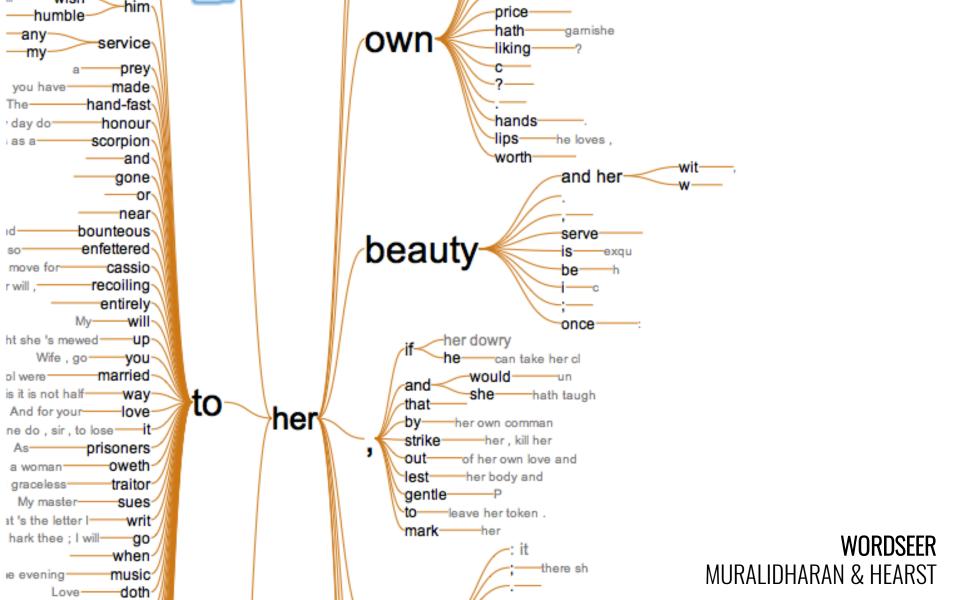
world, the love of the father is not in him.

children of god, when we love god, and keep his commandments

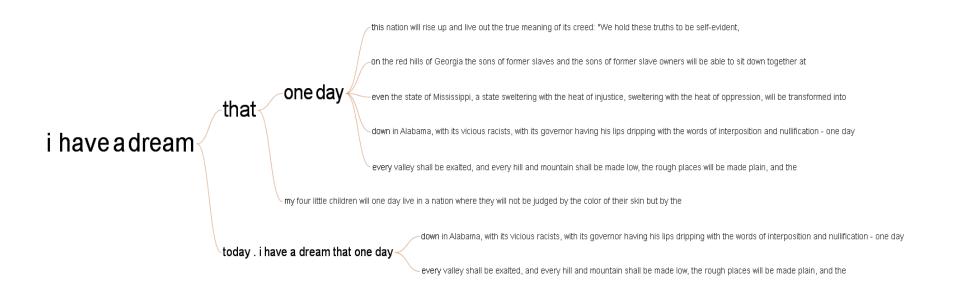
FILTER INFREQUENT RUNS







RECURRENT THEMES IN SPEECH



GLIMPSES OF STRUCTURE

- CONCORDANCES SHOW LOCAL, REPEATED STRUCTURE
- BUT WHAT ABOUT OTHER TYPES OF PATTERNS?

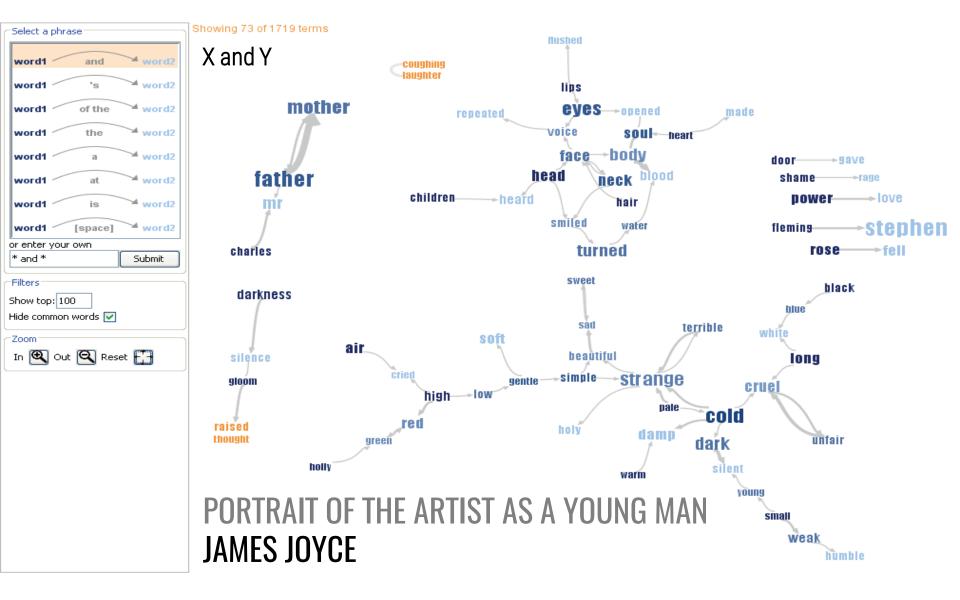
- FOR EXAMPLE
- LEXICAL: <A> at
- SYNTACTIC: <Noun> <Verb> <Object>

PHRASE NETS

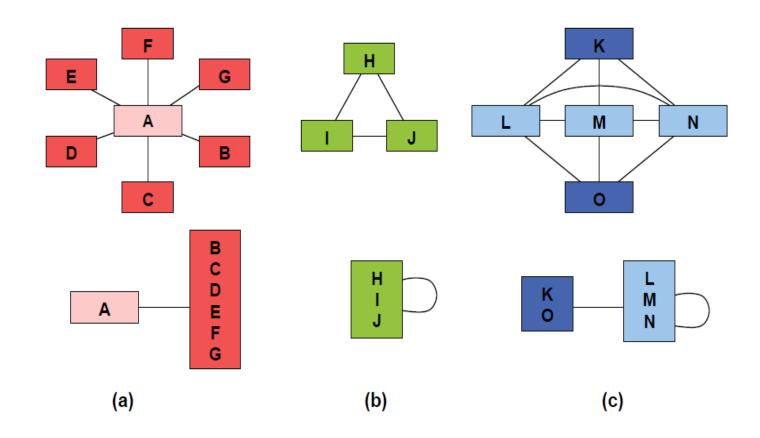
LOOK FOR SPECIFIC LINKING PATTERNS IN THE TEXT:

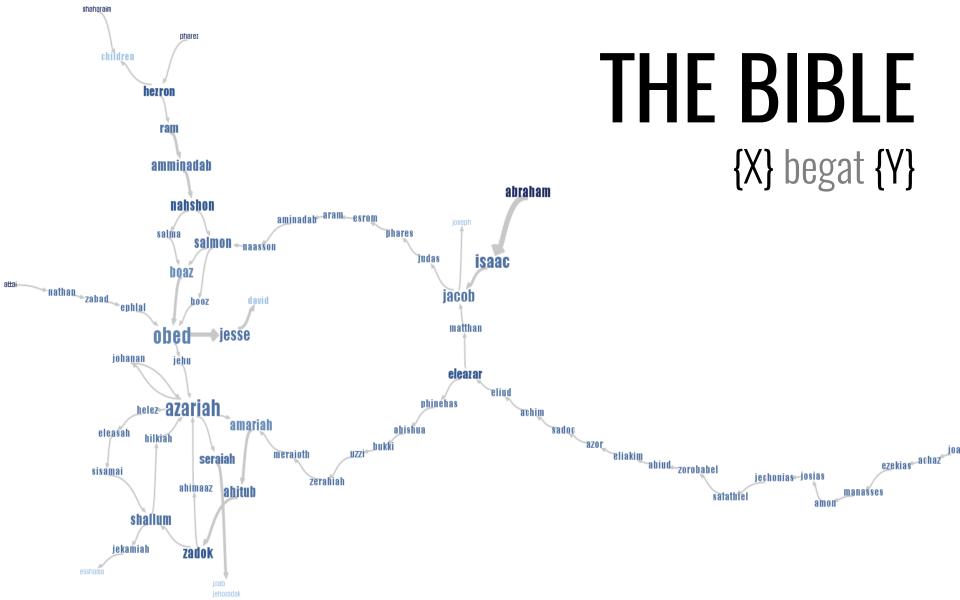
'A AND B', 'A AT B', 'A OF B', ETC COULD BE OUTPUT OF REGEXP OR PARSER

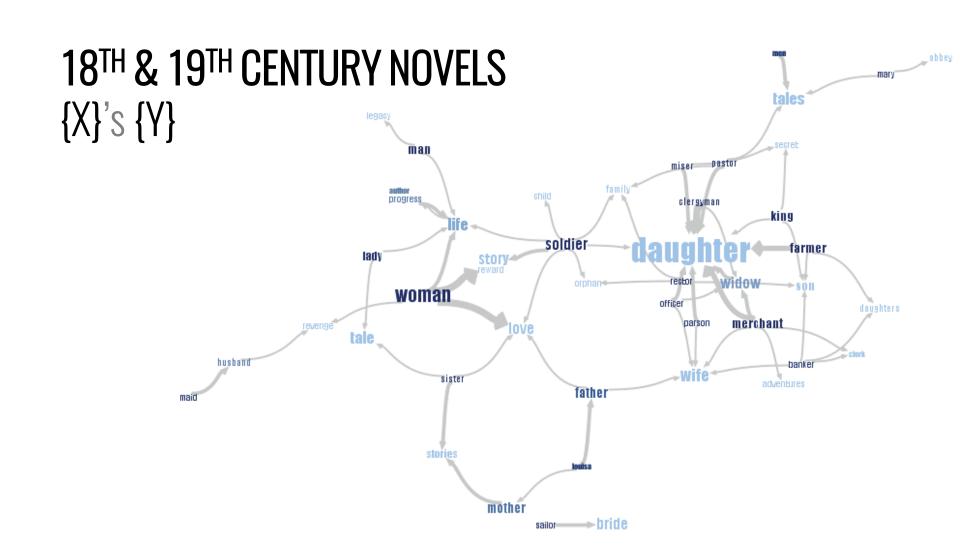
VISUALIZE EXTRACTED PATTERNS IN A NODE-LINK VIEW OCCURRENCES = NODE SIZE PATTERN POSITION = EDGE DIRECTION

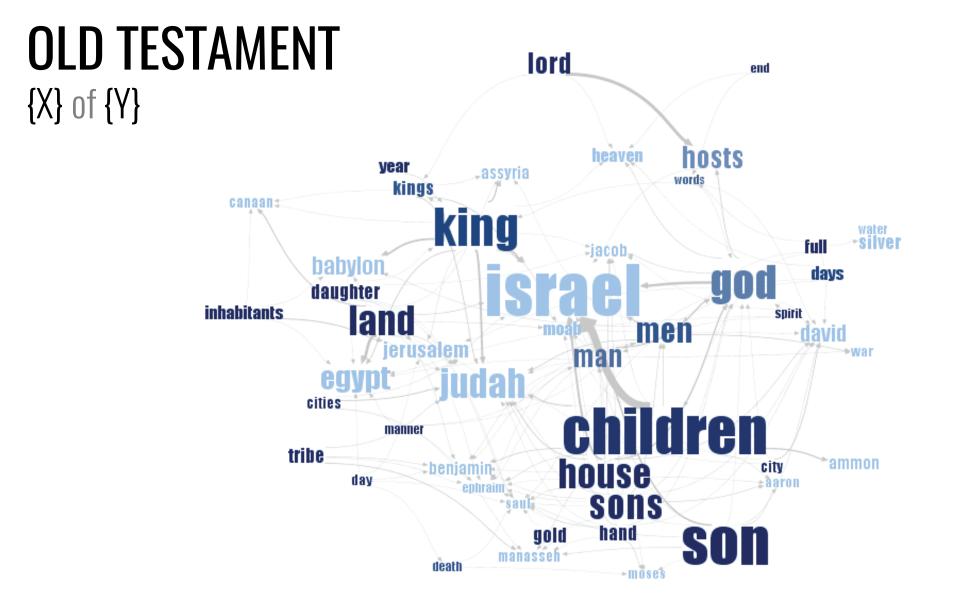


NODE GROUPING



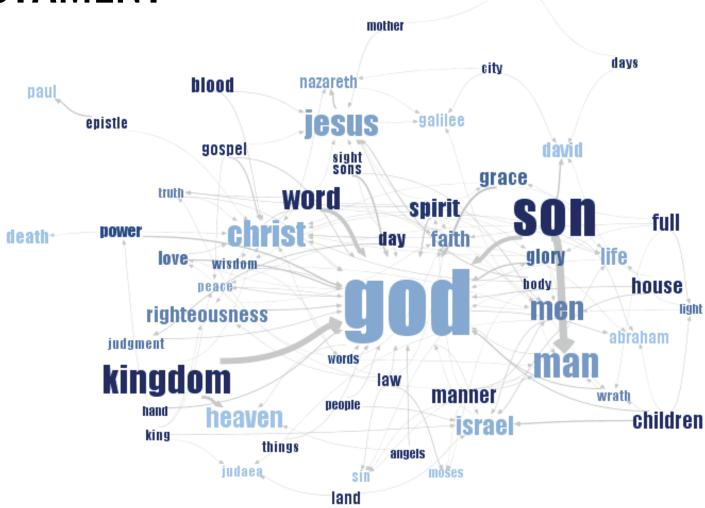






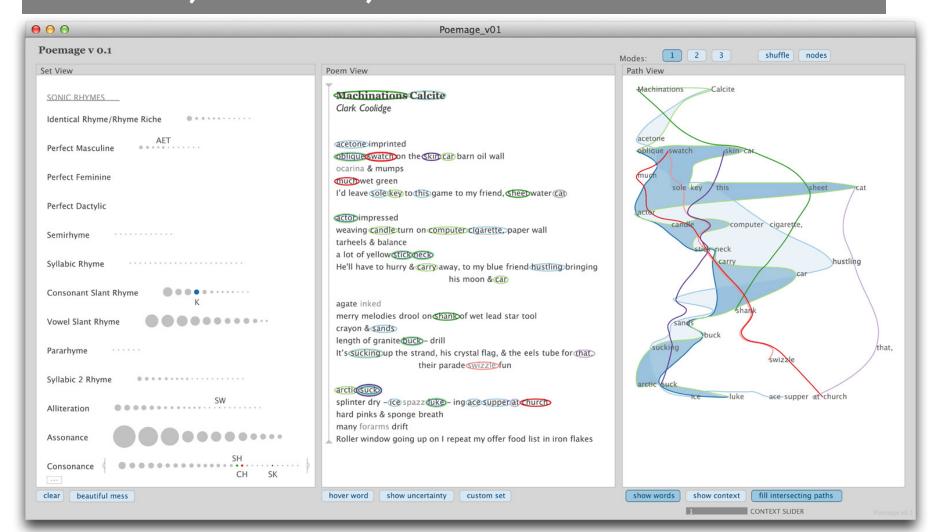
NEW TESTAMENT

 $\{X\}$ of $\{Y\}$



_iohn

RHYME, SPEECH, ETC. POEMAGE McCurdy et al. 2016



REVISIT YOUR SKETCHES?

TASK:

1) VISUALIZE THE MOST IMPORTANT CONTENT FROM A SINGLE PAPER.

ARE YOUR VISUALIZATION CHOICES **EFFECTIVE**?

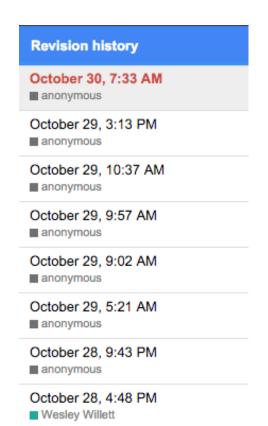
DOES THE VIS CAPTURE THE **LENGTH**, **FORM**, AND **POSITION** OF THE IMPORTANT CONTENT?

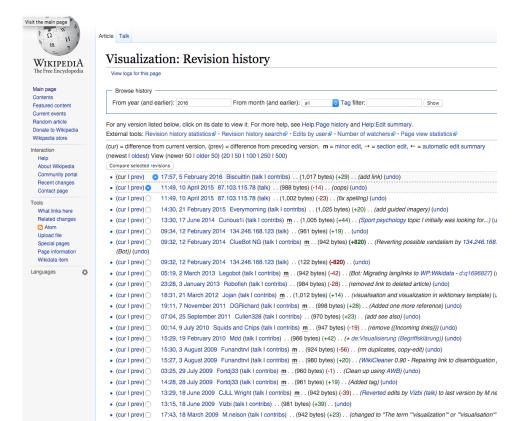
DO YOU SHOW OR CONNECT BACK TO THE **ORIGINAL TEXT**?

EVOLVING DOCUMENTS

VISUALIZING REVISION HISTORY

HOW TO DEPICT CONTRIBUTIONS AND CHANGES OVER TIME?





DIFF'

svn diff: sshconsole.js

Files Changed:

1. sshconsole.js: 1 change [1]

```
/home/toddw/src/sshconsole-read-only/content/sshconsole.js
                                                                               50 lines hidden [Expand]
51
           term = new VT100(80, 24, "term");
                                                                                                   term = new VT100(80, 24, "term");
52
                                                                                       52
          // term.debug = 1;
                                                                                                  // term.debug = 1;
53
           _term.curs_set(true, true, _term_box_element);
                                                                                                   term.curs set(true, true, term box element);
          term.noecho();
                                                                                                  term.noecho();
55
          // Replace the go getch function with our own, this is called
                                                                                                  // Replace the go getch function with our own, this is called
          // for every keypress that is passed through the terminal to the
                                                                                                  // for every keypress that is passed through the terminal to the
          // remote server. The character is already converted into the
                                                                                                  // remote server. The character is already converted into the
          // required VT100 character sequence(s).
                                                                                                  // required VT100 character sequence(s).
          VT100.go getch = function() {
                                                                                                  VT100.go_getch_ = function() {
              var vt = VT100.the vt;
                                                                                       61
                                                                                                      var vt = VT100.the vt ;
                                                                                       62
              if (vt === undefined) {
                                                                                                      if (vt === somevalue) {
                      return;
                                                                                                              return;
              var ch = vt.key buf .shift();
                                                                                                      var ch = vt.key buf .shift();
              //dump("go getch :: ch: '" + ch + "'\n");
              if (ch === undefined) {
                                                                                                      if (ch === undefined) {
                      return;
                                                                                                              return;
              if (vt.echo && ch.length == 1) {
                                                                                                      if (vt.echo && ch.length == 1) {
71
                      vt.addch(ch);
                                                                                                              vt.addch(ch);
                                                                                                              vt.refres();
                                                                                       72
72
73
              if ( ssh channel) {
                                                                                                      if ( ssh channel) {
74
                   ssh channel.sendStdin(ch);
                                                                                                          ssh channel.sendStdin(ch);
75
76
                                                                                       76
77
78
          var serverTextbox = document.getElementById("sshconsole server textbox");
                                                                                                  var serverTextbox = document.getElementById("sshconsole server textbox");
79
          var connectionText;
                                                                                                  var connectionText;
          if ('connectionText' in window.arguments[0]) {
                                                                                                  if ('connectionText' in window.arguments[0]) {
              connectionText = window.arguments[0].connectionText;
                                                                                                      connectionText = window.arguments[0].connectionText;
          } else {
                                                                                                  } else {
                                                                               174 lines hidden [Expand]
```



Abortion

(Revision as of 22:56 4 Jun 2003)

"Abortion," in its most commonly used sense, refers to the defiberate early termination of prepancy, resulting in the death of the embryo or featus. [1] Medically, the term also refers to the early termination of a pregnancy by natural causes ("spontaneous abortion" or miscarriage, which ends in S of all pregnances, susually within the first 13 weeks) or to the cessation of normal growth of a body part or organ, What follows is a discussion of the issues related to deliberate or "induced" abortion.

Methods

Depending on the stage of pregnancy an abortion is performed by a number of different methods. For the earliest terminations (before nine weeks or so) a chemical abortion is the usual method, the drug mifepristens is usually the only legal method although research has uncovered similar effects from methorizexate and misconorstol. Concurrent with chemical abortion and extending up until around the fifteenth week section-aspiration or vacuum abortion is the most common approach, replacing the more risky dilation, and curettage. (O. &. C). From the fifteenth week up until around the eighteenth week a surgical dilation, and execusion. (D. &. E) is used.

As the feture size increases other techniques must be used to secure abortion in the third trimination, premature expulsion of the feture can be induced with prostablandin, this can be coupled with injecting the aministic fluid with saline or ureal solution. Very late abortions can be brought about by the controversal instant dilation and extraction (D & X) or a histocrotimy abortion, similar to a caesarian section.

The controversy

The morality and legality of abortion is a large and important topic in applied ethics and is also discussed by legal scholars and religious people. Important facts about abortion are also researched by sociologists and historians.

Abortion has been common in most societies, although it has often been opposed by some institutionalized religions and governments. In 20th, sentury politics in the United States and Europe, abortion became commonly accepted by the end of the 20th century. Additionally, abortion is legal and accepted in China. India and other populous countries. The Catholic Church remains opposed to the procedure, however, and in other countries, notably the United States and the (predominantly Catholic) Republic of Ireland, the controversy is still extremely active, to the extent that even the names of the respective positions are subject to head debate. While those on both sides of the argument are generally peaceful, if heated, in their advocacy of their positions, the debate is sometimes characterized by violence. Though true of both sides, this is more marked on the side of those opposed to abortion, because of what they see as the gravity and urgency of their views.

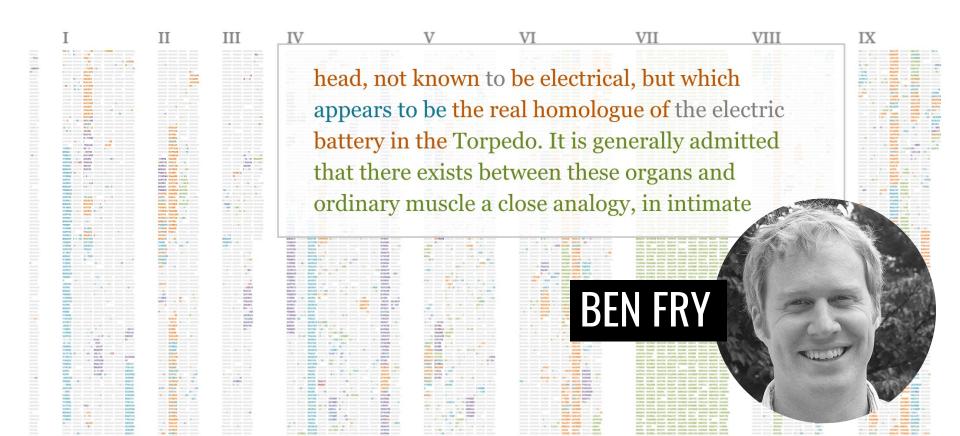
The central question

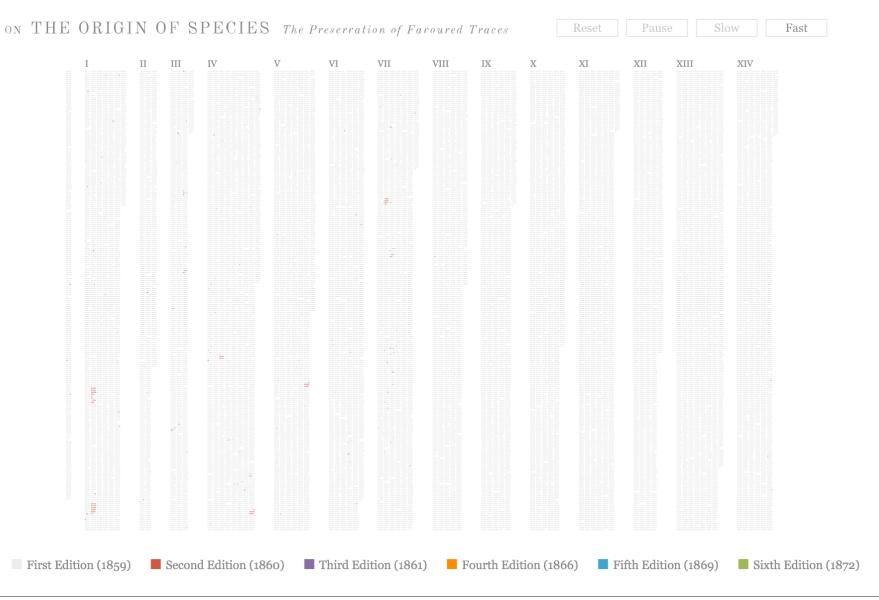
The central question in the abortion debate is a clash of presumed or perceived rights. On the one hand, is a fetus (sometimes called the "unborn" by pro-life/anti-abortion advocates) a human being with a right to life, and if so, at what point in the

ANIMATED TRACES

fathom.info/traces

ON THE ORIGIN OF SPECIES The Preservation of Favoured Traces





DIFFAMATION

Using Text Animated Transitions to support Navigation in Document Histories



Fanny Chevalier



Pierre Dragicevic



Anastasia Bezerianos



Jean-Daniel Fekete

Shortest Edit Path

Edit War

VISUALIZING DOCUMENT COLLECTIONS

SKETCHING: VISUALIZE

IMAGINE YOU HAVE 20 YEARS OF SCIENTIFIC PAPER ABSTRACTS:

YEAR

CONFERENCE

AUTHORS

TITI F

PAPER TYPE

KEYWORDS

REFERENCES

ABSTRACT TEXT

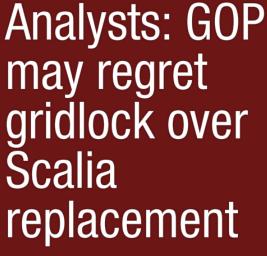
TASK:

1) VISUALIZE THE MOST IMPORTANT CONTENT FROM A SINGLE PAPER.

2) VISUALIZE HOW SIMILAR PAPERS FROM EACH CONFERENCE ARE TO PAPERS FROM OTHER CONFERENCES.

GROUPS OF 3

(~10 MINUTES)



Update: Uber driver arrested in Michigan rampage that killed 6

Boris Johnson backs EU exit: London mayor confirms support for Brexit

'A multifaceted catastrophe': Turkey has 'so alienated everyone Blasts rock Syrian city of Homs, killing at least 32 Palestinians struggle to define

those who attack Israelis

Canada, USA

renew rivalry

in CONCACAF

Malaysia, south-IS rejected North east Asia nations orea peace warned of terror attacks alks offer before ast nuclear test

'Deadpool' dominates again with \$55 million in 2nd week

Judge blocks attempt to halt deposition of Bill Cosby's wife

Taylor Swift donates \$250K to help Kesha's legal battle

Scientists at Brock

studying Zika to see if

Canadian mosquitoes can spread the virus

Highlights from the USC report on

gridlock over

Samsung, LG unveil new devices in bid for smartphone recovery

LG Unveils the LG G5, Its First Modular Smartphone [Video]

LG G5 vs LG V10: first look

short in closest Daytona 500

Truex comes up a few inches

Chan wins Four Continents

figure-skating championship

Raceline Radio

Program Guide:

February 21,

2016

Years later, ex-Raptor Vince Carter's still

final

Canadian women earn historic 19-10 rugby

Leafs get set for a busy draft with Matthias trade

SPRING TRAINING Blue Jays' focus at

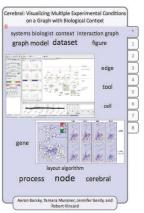
2016 camp is on 2017

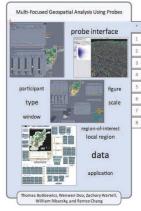
Miller puts an end to

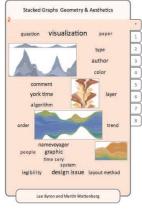
newsmap.jp

DOCUMENT CARDS

SMALL MULTIPLES FOR DOCUMENTS









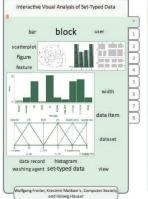




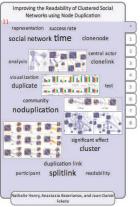




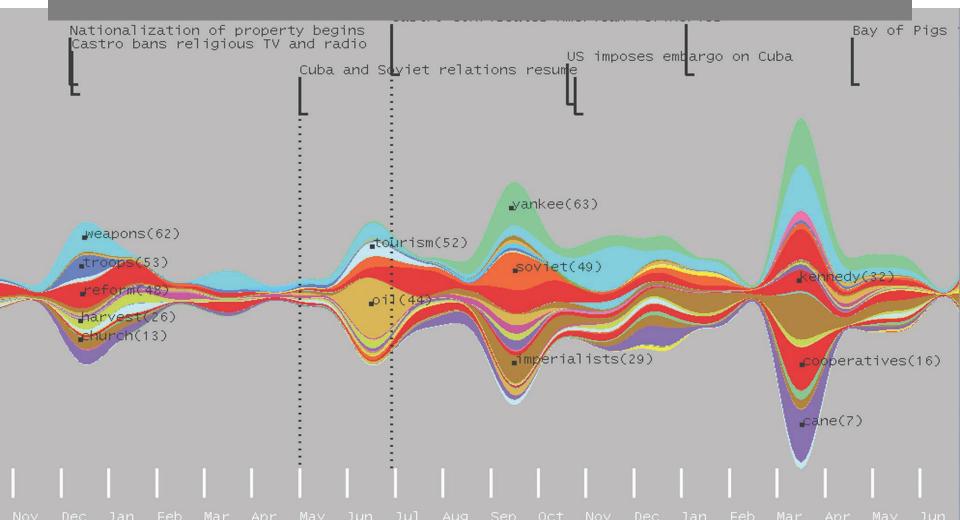




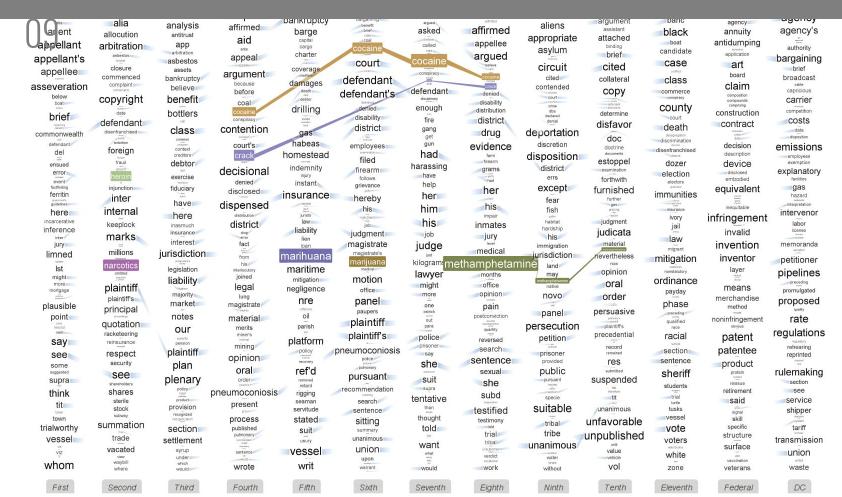




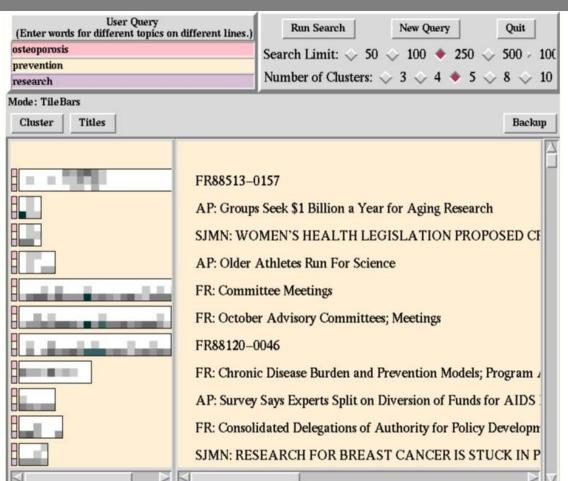
THEMERIVER HAVRE ET AL 1999



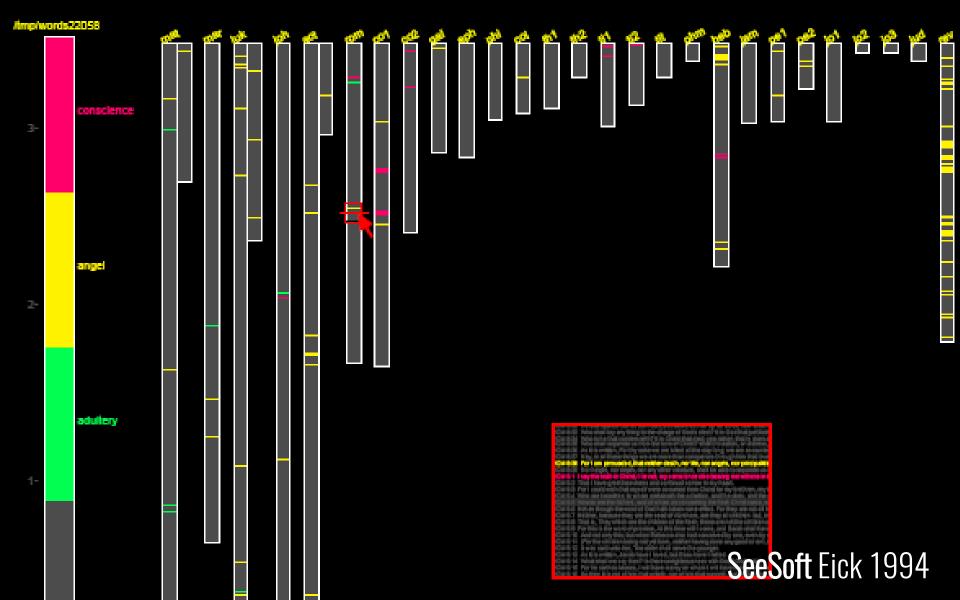
PARALLEL TAG CLOUDS



SUPPORTING SEARCH

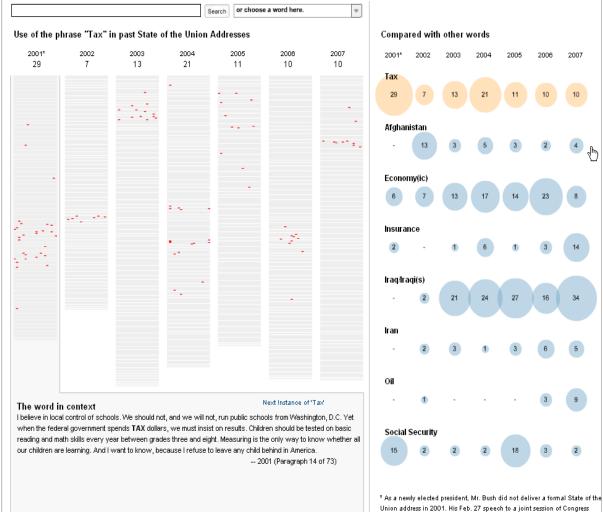


TileBars Hearst 1999



The 2007 State of the Union Address

Over the years, President Bush's State of the Union address has averaged almost 5,000 words each, meaning the the President has delivered over 34,000 words. Some words appear frequently while others appear only sporadically. Use the tools below to analyze what Mr. Bush has said.



Ben Werschkul/The New York Times

was analogous to the State of the Union, but without the title.

NAMED ENTITY RECOGNITION

IDENTIFY AND CLASSIFY NAMED ENTITIES IN TEXT:

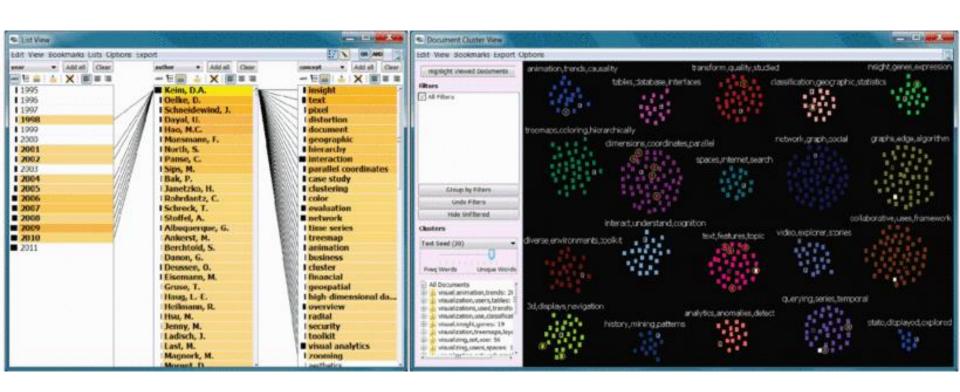
JOHN SMITH IS A PERSON SOVIET UNION IS A COUNTRY 2500 UNIVERSITY DR IS AN

ADDRESS

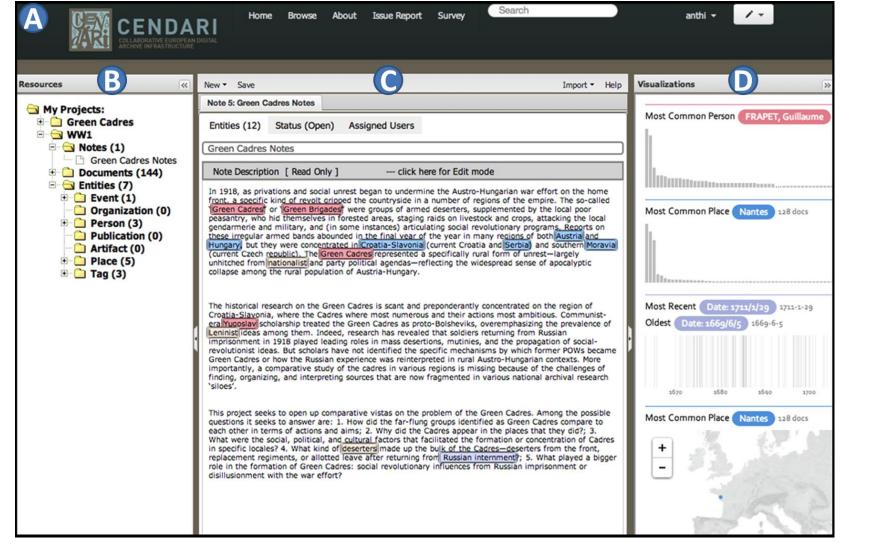
(555) 867-5309 IS A PHONE NUMBER

ENTITY RELATIONS: HOW DO THE ENTITIES RELATE?

DO THEY CO-OCCUR IN A DOCUMENT? IN A SENTENCE?



JIGSAW



CENDARI NOTE-TAKING ENVIRONMENT 2015

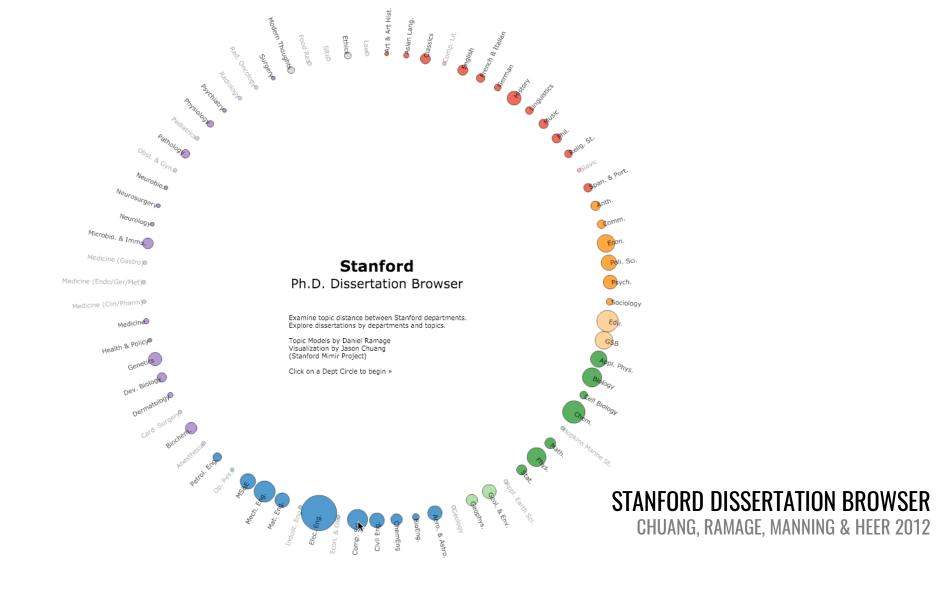
DOCUMENT SIMILARITY & CLUSTERING

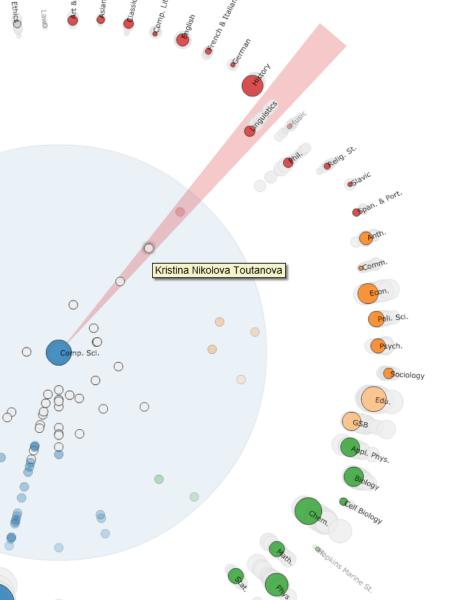
COMPUTE SIMILARITY BETWEEN DOCUMENTS BASED ON THE WORDS THEY SHARE

■ TF-IDF (TERM FREQUENCY-INVERSE DOCUMENT FREQUENCY) IS COMMON

TOPIC MODELING APPROACHES

- ASSUME DOCUMENTS ARE A MIXTURE OF TOPICS
- TOPICS ARE (ROUGHLY) A SET OF CO-OCCURRING TERMS
- LATENT SEMANTIC ANALYSIS (LSA): REDUCE TERM MATRIX
- MANY, MANY APPROACHES EXIST





Effective statistical models for syntactic and semantic disambiguation

Student: Kristina Nikolova Toutanova Advisor: Christopher D. Manning

Computer Science (2005)

Keywords: Syntactic, Semantic, Tree kernels, Parsing

Abstract:

This thesis focuses on building effective statistical models for disambiguation of sophisticated syntactic and semantic natural language (NL) structures. We advance the state of the art in several domains by (i) choosing representations that encode domain knowledge more effectively and (ii) developing machine learning algorithms that deal with the specific properties of NL disambiguation tasks--sparsity of training data and large, structured spaces of hidden labels. For the task of syntactic disambiguation, we propose a novel representation of parse trees that connects the words of the sentence with the hidden syntactic structure in a direct way. Experimental evaluation on parse selection for a Head Driven Phrase Structure Grammar shows the new representation achieves superior performance compared to previous models. For the task of disambiguating the semantic role structure of verbs, we build a more accurate model, which captures the knowledge that the semantic frame of a verb is a joint structure with strong dependencies between arguments. We achieve this using a Conditional Random Field without Markov independence assumptions on the sequence of semantic role labels. To address the sparsity problem in machine learning for NL, we develop a method for incorporating many additional sources of information, using Markov chains in the space of words. The Markov chain framework makes it possible to combine multiple knowledge sources, to learn how much to trust each of them, and to chain inferences together. It achieves large gains in the task of disambiguating prepositional phrase attachments.

STANFORD DISSERTATION BROWSER

CHUANG, RAMAGE, MANNING & HEER 2012

WARNING

OFTEN, TEXT VISUALIZATIONS DO NOT REPRESENT TEXT DIRECTLY, BUT THEY REPRESENT A MODEL

WORD COUNTS, WORD SEQUENCES, CLUSTERS, ETC.

ASK:

CAN YOU INTERPRET THE VISUALIZATION?

DOES THE MODEL ACCURATELY REPRESENT THE ORIGINAL TEXT?

LESSONS FOR TEXT VISUALIZATION

SHOW SOURCE TEXT (OR PROVIDE ACCESS TO IT)
WHERE POSSIBLE, USE VISUALIZATION AS INDEX INTO DOCUMENTS

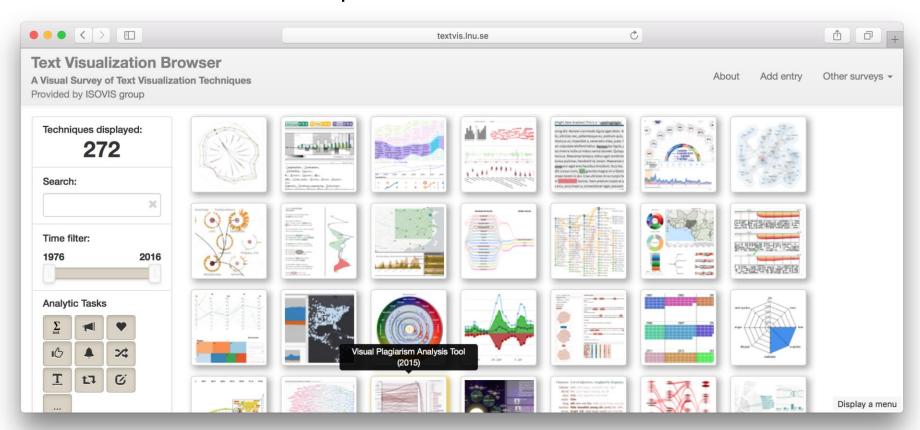
GROUP DOCUMENTS IN MEANINGFUL WAYS

WILL VIEWERS UNDERSTAND THE CLUSTERS?

WHERE POSSIBLE **USE TEXT TO REPRESENT TEXT**

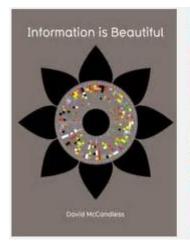
HUNDREDS OF TOOLS & TECHNIQUES FOR TEXT AT

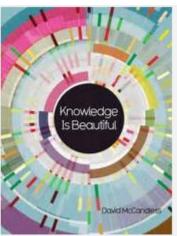
http://textvis.lnu.se/



HTTPS://YOUTU.BE/HOEX1IU57IW?T=420

Getting to know David McCandless





HTTP://WWW.WEFEELFINE.ORG/

QUESTIONS?

ACKNOWLEDGEMENTS

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- Christopher Collins (University of Ontario Institute of Technology)
- Wesley Willett (University of Calgary)