Making Sense of Web Data with Natural Language Processing

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About Me

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- fluquid.com
 - Sales & Client Intelligence
 - Intelligent Lead Generation
 - Large-scale web crawls
 - Gathering and Enriching Web Data
- webdata.org
 - Share Libraries and Best Practices
 - Bring Data Scientists and SME Companies together
 - ForDevelopers
 - AwesomeAvailableDatasets
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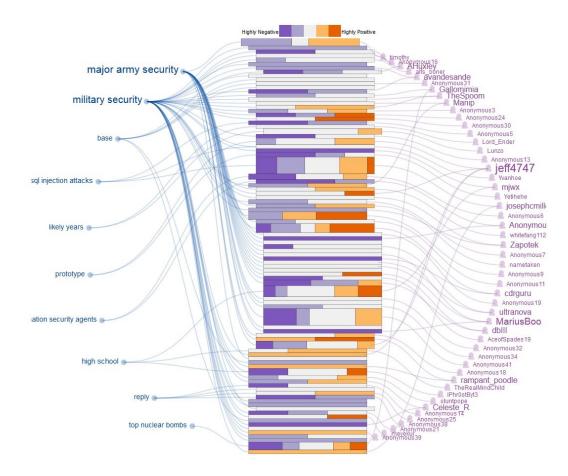
Data is Noisy

- Data is noisy (typos, free text, etc.) ("Mnuich", "Munich", "munich")
- Data can vary syntactically ("12.00", 12.00, 12)
- Many ways to represent the same entity ("Munich", "München", "Muenchen", "Munique", "48.1351° N, 11.5820° E", "zip 80331–81929", "[ˈmʏnçn̩]", "Minga", "慕尼黑")
- Entity representations are ambiguous
 - <Munich City, Germany>
 - <Munich County, Germany>
 - <Munich, North Dakota>
- Wikipedia disambiguation



Natural Language Processing

- 1. Content Extraction
- 2. Parsing
- 3. Named Entity Extraction,
- 4. Topic Modelling
- 5. Sentiment Analysis





1) Content Extraction

- Challenge:
 Given a document,
 extract the main text information
 as plaintext
- Libraries
 - html-text
 - <u>boilerpipe</u> (java)
 - dragnet
 - apache tika (java; supports many formats)
- Example Readability





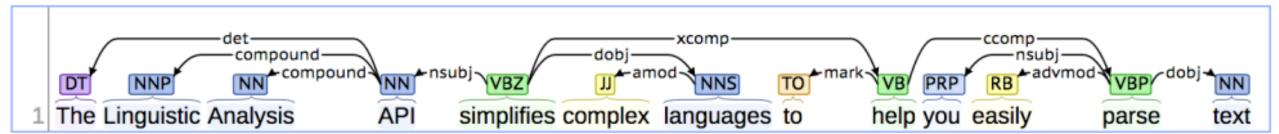
2) Parsing

- Spacy 2 is awesome!
 - Sentence segmentation
 - Word segmentation
 - Lemmatization/stemming
 - Parsing
 - POS (part of speech)
 - Word vectors
 - Word/sentence similarity
 - etc.

- Textacy
 - Extends spacy functionality
- syntaxnet
 - Parser and language understanding engine developed by Google
 - For more advanced use cases

Enhanced Dependencies:

Image:https://stanfordnlp.github.io/CoreNLP/images/Cate-Blanchett.png



3) Named Entity Extraction

- Entities:
 persons, organizations, locations, date, time, money,
 email, social media, postal address, etc.
- NER, Disambiguation
 - spacy basic entity extraction
 - <u>stanbol</u> pretty good for "production use"
 - <u>dbpedia spotlight</u> between stanbol and AIDA
 - AIDA very good, but slow
- Normalization
 - <u>cleanco</u> companies
 - <u>probablepeople</u> person names
 - <u>python-phonenumbers</u> international phone numbers
 - <u>libpostal</u> postal addresses
- webstruct train your own NER with annotated training data

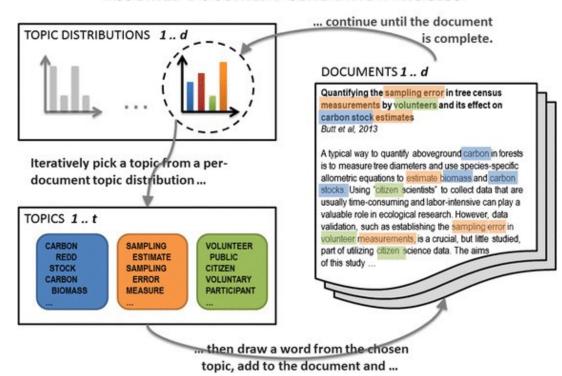




4) Topic Modelling

- Goal: Dimensionality Reduction from 50k+dimensional token space to "topic" manifold
- Assumption: Every document covers several different "topics"
- A topic is comprised of words that often co-occur
- Approach: Analyze which words co-occur more frequently with each other than with other words
- Can be used as a basis for clustering, similarity, etc.
- Libraries
 - gensim LDA
 - sklearn NMF
- Demo

ASSUMED DOCUMENT GENERATION PROCESS





5) Sentiment Analysis

- Identify what sentiment an expression carries
 - Polarity, Subjectivity
 - Paragraph, Sentence, Entity
- Challenges:
 - Generally messy and often does not produce great results
 - Sarcasm, Irony, Context
 - Mixed sentiments in any single statement
- Libraries
 - vaderSentiment
 - <u>twitter-sent-dnn</u>
- Examples
 - <u>cryptocurrencies</u>
 - <u>twitter "performance review" tweets</u>





Metadata

- Use pre-structured information from web data where available
- Formats
 - Metadata (schema.org)
 - Microdata (vcard)
 - json-ld
 - OpenGraph
 - Twitter Card
- Libraries
 - Extruct
 - Apache Any23 (java)





Miscellaneous

- Language Detection
 - cld2-cffi
- Find many possible terms in text
 - <u>pyahocorasick</u>
- Structured Data Extraction
 - Pydepta
 - <u>Demo</u>
- Unicode Normalization
 - <u>unidecode</u>





Questions?

- Content Extraction in R
 - boilerpipeR
- Wordpress Plugin Scanner
 - sorry, it's not open-source yet; but I will open-source it soon at github.com/fluquid
- Extract Bibliography from Academic Papers
 - grobid (GeneRation Of BIbliographic Data)
 - pdfextract
 - <u>CERMINE</u>
- Find similar skills, capabilities
 - gensim word2vec
 - spacy even comes with <u>semantic sentence similarity</u>;)

