Introduction to Natural Language Processing

Document Classification

Problem Motivation

- You're Google News, and you want to group news articles by topic.
- You're a legal tech firm, and you need to sift through 100,000 pages of legal documents to find the relevant ones.

Overview of Approach

- Compile documents
- Featurize them
- Compare their features

Simple Sample Problem

- 2 documents: "blue house" and "red house"
- Could featurize based on word counts
 - \circ "blue house" \rightarrow (red, blue, house) = (0, 1, 1)
 - \circ "red house" \rightarrow (red, blue, house) = (1, 0, 1)

Bag of Words

 A document represented as a vector of word counts is called a "bag of words"

- "blue house" → (red, blue, house) = (0, 1, 1)
- "red house" \rightarrow (red, blue, house) = (1, 0, 1)
- "red red house" \rightarrow (red, blue, house) = (2, 0, 1)

Comparing the Features: Cosine Similarity

similarity =
$$cos(\theta) = \frac{A \cdot B}{\|A\| \|B\|}$$

"red house" vs "blue house"
$$\rightarrow$$
 (1,0,1) vs (0, 1, 1) \rightarrow similarity = .5

Is there a better way to featurize?

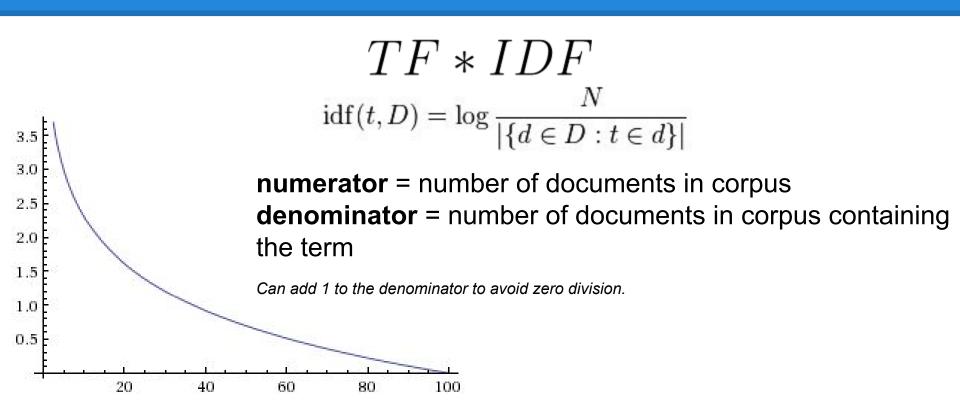
- Bag of words is naive--just word counts!
- Every word has equal weighting
 - "the" and "data" have different predictive power
- Can adjust word counts based on their frequency in the corpus

TF-IDF

$$TF*(?)$$

- How to adjust the term frequency?
- Words found in only one document should have highest weighting
- Words found in every document should have lowest weighting

IDF - Inverse Document Frequency



Feature Engineering for Text

- Tokenize
- Remove stop words
- Stemming
 - e.g. "walked" → "walk"
- Lemmatization (better than stemming)
 - e.g. "people" → "person"
- N-grams / skip grams

More Advanced NLP Problem Types

Sentiment analysis

http://nlp.stanford.edu/sentiment/index.html

Machine translation

https://medium.com/s-c-a-l-e/how-baidu-mastered-mandarin-with-deep-learning-and-lots-of-data-1d94032564a5

Probabilistic parsing