TF-IDF Representation

Dot Product, Norm, and Cosine Review?

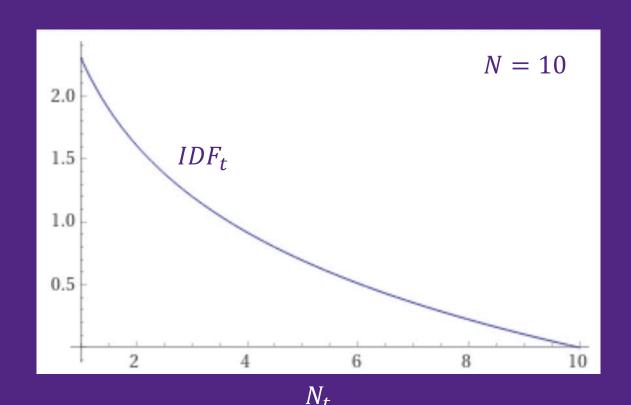
TF-IDF

- Term Frequency Inverse Document Frequency
- Different vector representation for documents
- Replaces BoW counts to reflect term "importance" relative to the corpus.
 - Words that are less widespread in the corpus get more weight.
- Many variants

Inverse Document Frequency

- For term *t*,
 - Let N be number of documents
 - Let N_t be number of documents containing term t

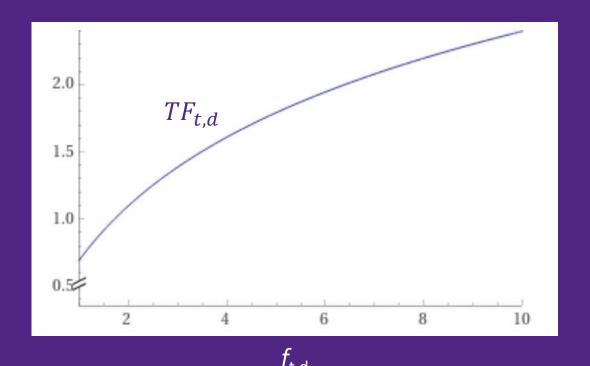
•
$$IDF_t = log\left(\frac{N}{N_t}\right)$$



Term Frequency

- Higher frequency implies higher importance
 - Empirically, a diminishing return is helpful
- ullet Let $f_{t,d}$ be the frequency of term t in document d

• $TF_{t,d} = \log(1 + f_{t,d})$



TF-IDF

•
$$TFIDF_{t,d} = log(1 + f_{t,d}) \times log(\frac{N}{N_t})$$

The TF-IDF "Vector model" Dense Representation

| DocID | first | hurlyburly | in | thunder | witch | witchcraft | witches | witching | |
|-------|-------|------------|-------|---------|-------|------------|---------|----------|-----|
| | | | | | | | | | |
| 1 | 0.931 | 0 | 0.931 | 0.388 | 0.931 | 0.931 | 0 | 0 | ••• |
| 4 | 0 | 0 | 0 | 0 | 0 | 0 | 1.376 | 0 | ••• |
| 5 | 0 | 0 | 0 | 0.388 | 0 | 0.587 | 0 | 0 | |
| 8 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1.349 | |
| 9 | 0 | 0.868 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 22 | 1.175 | 0.868 | 0.931 | 0.615 | 1.364 | 0 | 1.376 | 0 | ••• |
| 37 | 0.587 | 0 | 0.587 | 0.388 | 0.587 | 0.587 | 0 | 0 | ••• |
| | ••• | ••• | ••• | ••• | ••• | ••• | ••• | ••• | ••• |

The TF-IDF "Vector model" Sparse Representation

| DocID | Words | | | |
|-------|--|--|--|--|
| 1 | first:0.931, in:0.931, thunder:0.388, witch:0.931, witchcraft:0.931 | | | |
| 4 | witches:1.376 | | | |
| 5 | thunder:0.388, witchcraft:0.587 | | | |
| 8 | witching:1.349 | | | |
| 9 | hurlyburly:0.868 | | | |
| 22 | first:1.175, hurlyburly:0.868, in:0.931, thunder:0.615, witch:1.364, witches:1.376 | | | |
| 37 | first:0.587, in:0.587, thunder:0.388, witch:0.587, witchcraft:0.587 | | | |
| | ··· | | | |

Similarity – TFIDF Cosine measure

| DocID | Words | Similarity to {baseball:0.13, season:0.13, opener:1.24} | | | | |
|-------|---|---|--|--|--|--|
| 1 | baseball:0.44, season:0.13, opener:1.24 | 0.972 | | | | |
| 2 | baseball:0.44, season:0.33 | 0.141 | | | | |
| 6 | season:0.13 | 0.101 | | | | |
| 7 | baseball:0.44 | 0.101 | | | | |
| 10 | baseball:0.44, season:0.25 | 0.138 | | | | |
| 35 | baseball:0.44, season:0.20 | 0.134 | | | | |
| | | | | | | |

Generalized Similarity Measures

BM-25

General Similarity Measures

 "Tune" the ideas of bag-of-words, TFIDF, dot product/cosine to perform well for information retrieval

Queries and documents are not treated the same.
(Similarity from query-to-document not same as from document-to-query.)

Default in Lucene

BM25

Lucene default similarity function, related to TF-IDF

•
$$BM25_{d,q} = \sum_{t} q_t \frac{(k+1)f_{t,d}}{f_{t,d} + k\left((1-b) + b\frac{|d|}{avg\ doc\ length}\right)} IDF_t$$

- The numbers k and b are "tuning parameters"
- q_t is frequency of term t in the query
- |d| is document length; $avg\ doc\ length$ is average document length

BM25

• Lucene defaults: k = 1.2, b = 0.75

•
$$BM25_{d,q} = \sum_{t} q_{t} \frac{2.2f_{t,d}}{f_{t,d} + 0.3 + 0.9 \frac{|d|}{avg \ doc \ length}} IDF_{t}$$

- q_t is frequency of term t in the query
- |d| is document length; $avg\ doc\ length$ is average document length
- Think:
 - What happens as IDF goes up?
 - What happens as $f_{t,d}$ goes up?
 - What happens as |d| goes up?

Summary – Document Representation and Retrieval

- Term search and Boolean Queries
- Similarity-based Search
 - Dot product
 - Cosine
- Representations
 - Bag of Words (counts)
 - TFIDF
- Generalized similarity measures
 - BM25

Linear Algebra

- If you need to refresh the basics of linear algebra:
 - Multiplication of vectors and matrices
 - Transpose
 - Matrix Inverse
- http://www.cs.cmu.edu/ ~zkolter/course/15-884/ linalg-review.pdf



I'll review these with examples.