

3-3: TFIDF and More!

Introduction to Recommender Systems

The search problem ...

- Why do primitive search engines fail?
- What would a primitive search engine do?
 - Return all documents that contain search terms?
 - More frequent occurrence ranked higher?
- At a minimum, need to consider two factors
 - Term frequency may be significant
 - Not all terms equally relevant
- Actually, much harder than this, more later ...

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Learning Objectives

- To understand the problem that requires a weighting for search or filtering
- To understand TFIDF weighting in detail, and how it is used in both search and filtering
- To understand the range of variants and alternatives to TFIDF
- To appreciate the similarities and differences between content filtering and search

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TFIDF weighting

- Term Frequency * Inverse Document Frequency
- Term Frequency =
 - Number of occurrences of a term in the document (can be a simple count)
- Inverse Document Frequency =
 - How few documents contain this term
 - Typically
$$\log(\text{\#documents} / \text{\#documents with term})$$

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What does TFIDF do?

- Automatic demotion of stopwords, common terms
- Promotes core terms over incidental ones

But where does it fail?

- If core term/concept isn't actually used (much) in document (e.g., legal contracts)
- Poor searches (other techniques for that)

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Variants and Alternatives

- Some applications use variants on TF
 - 0/1 boolean frequencies (occurs above threshold)
 - Logarithmic frequencies ($\log(tf+1)$)
 - Normalized frequency (divide by document length)
- BM25 (aka Okapi BM25) is a ranking function used by search engines:
 - Includes frequency in query, in document, number of documents, length
 - Variants with different weights: BM11, BM15, ...

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How does TFIDF apply to CBF?

- TFIDF concept can be used to create a profile of a document/object
 - A movie could be described as a weighted vector of its tags (details next lecture)
- These TFIDF profiles can be combined with ratings to create user profiles, and then matched against future documents

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Actually much harder, as we said

- Phrases and n-grams
 - “computer science” != “computer” and “science”
 - Adjacency
- Significance in Documents
 - Titles, headings, ...
- General Document Authority
 - Pagerank and similar approaches
- Implied Content
 - Links, usage ...

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Take-Away and Moving Forward

- You should
 - Understand TFIDF and why it is needed
 - Also understand its limitations
- Next
 - Building and applying content profiles

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