

Chapter 19: Information Retrieval

Database System Concepts

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Chapter 19: Information Retrieval

- Relevance Ranking Using Terms
- Relevance Using Hyperlinks
- Synonyms., Homonyms, and Ontologies
- Indexing of Documents
- Measuring Retrieval Effectiveness
- Web Search Engines
- Information Retrieval and Structured Data
- Directories





Information Retrieval Systems

- Information retrieval (IR) systems use a simpler data model than database systems
 - Information organized as a collection of documents
 - Documents are unstructured, no schema
- Information retrieval locates relevant documents, on the basis of user input such as keywords or example documents
 - e.g., find documents containing the words "database systems"
- Can be used even on textual descriptions provided with non-textual data such as images
- Web search engines are the most familiar example of IR systems





Information Retrieval Systems (Cont.)

- Differences from database systems
 - IR systems don't deal with transactional updates (including concurrency control and recovery)
 - Database systems deal with structured data, with schemas that define the data organization
 - IR systems deal with some querying issues not generally addressed by database systems
 - Approximate searching by keywords
 - Ranking of retrieved answers by estimated degree of relevance





Keyword Search

- In full text retrieval, all the words in each document are considered to be keywords.
 - We use the word term to refer to the words in a document
- Information-retrieval systems typically allow query expressions formed using keywords and the logical connectives and, or, and not
 - Ands are implicit, even if not explicitly specified
- Ranking of documents on the basis of estimated relevance to a query is critical
 - Relevance ranking is based on factors such as
 - Term frequency
 - Frequency of occurrence of query keyword in document
 - Inverse document frequency
 - How many documents the query keyword occurs in
 - Fewer

 give more importance to keyword
 - Hyperlinks to documents
 - More links to a document → document is more important





Relevance Ranking Using Terms

- TF-IDF (Term frequency/Inverse Document frequency) ranking:
 - Let n(d) = number of terms in the document d
 - n(d, t) = number of occurrences of term t in the document d.
 - Relevance of a document d to a term t

$$TF(d, t) = log \left(1 + \frac{n(d, t)}{n(d)}\right)$$

- The log factor is to avoid excessive weight to frequent terms
- Relevance of document to query Q

$$r(d, Q) = \sum_{t \in Q} \frac{TF(d, t)}{n(t)}$$





Relevance Ranking Using Terms (Cont.)

- Most systems add to the above model
 - Words that occur in title, author list, section headings, etc. are given greater importance
 - Words whose first occurrence is late in the document are given lower importance
 - Very common words such as "a", "an", "the", "it" etc are eliminated
 - Called stop words
 - Proximity: if keywords in query occur close together in the document, the document has higher importance than if they occur far apart
- Documents are returned in decreasing order of relevance score
 - Usually only top few documents are returned, not all





Similarity Based Retrieval

- Similarity based retrieval retrieve documents similar to a given document
 - Similarity may be defined on the basis of common words
 - E.g. find *k* terms in A with highest *TF* (*d*, *t*) / *n* (*t*) and use these terms to find relevance of other documents.
- Relevance feedback: Similarity can be used to refine answer set to keyword query
 - User selects a few relevant documents from those retrieved by keyword query, and system finds other documents similar to these
- Vector space model: define an *n*-dimensional space, where *n* is the number of words in the document set.
 - Vector for document d goes from origin to a point whose i the coordinate is TF (d,t) / n (t)
 - The cosine of the angle between the vectors of two documents is used as a measure of their similarity.





Relevance Using Hyperlinks

- Number of documents relevant to a query can be enormous if only term frequencies are taken into account
- Using term frequencies makes "spamming" easy
 - E.g. a travel agency can add many occurrences of the words "travel" to its page to make its rank very high
- Most of the time people are looking for pages from popular sites
- Idea: use popularity of Web site (e.g. how many people visit it) to rank site pages that match given keywords
- Problem: hard to find actual popularity of site
 - Solution: next slide





Relevance Using Hyperlinks (Cont.)

- Solution: use number of hyperlinks to a site as a measure of the popularity or prestige of the site
 - Count only one hyperlink from each site (why? see previous slide)
 - Popularity measure is for site, not for individual page
 - But, most hyperlinks are to root of site
 - Also, concept of "site" difficult to define since a URL prefix like cs.yale.edu contains many unrelated pages of varying popularity
- Refinements
 - When computing prestige based on links to a site, give more weight to links from sites that themselves have higher prestige
 - Definition is circular
 - Set up and solve system of simultaneous linear equations
 - Above idea is basis of the Google PageRank ranking mechanism





Relevance Using Hyperlinks (Cont.)

- Connections to social networking theories that ranked prestige of people
 - E.g. the president of the U.S.A has a high prestige since many people know him
 - Someone known by multiple prestigious people has high prestige
- Hub and authority based ranking
 - A hub is a page that stores links to many pages (on a topic)
 - An authority is a page that contains actual information on a topic
 - Each page gets a hub prestige based on prestige of authorities that it points to
 - Each page gets an authority prestige based on prestige of hubs that point to it
 - Again, prestige definitions are cyclic, and can be got by solving linear equations
 - Use authority prestige when ranking answers to a query





Synonyms and Homonyms

Synonyms

- E.g. document: "motorcycle repair", query: "motorcycle maintenance"
 - need to realize that "maintenance" and "repair" are synonyms
- System can extend query as "motorcycle and (repair or maintenance)"
- Homonyms
 - E.g. "object" has different meanings as noun/verb
 - Can disambiguate meanings (to some extent) from the context
- Extending queries automatically using synonyms can be problematic
 - Need to understand intended meaning in order to infer synonyms
 - Or verify synonyms with user
 - Synonyms may have other meanings as well





Concept-Based Querying

- Approach
 - For each word, determine the concept it represents from context
 - Use one or more ontologies:
 - Hierarchical structure showing relationship between concepts
 - E.g.: the ISA relationship that we saw in the E-R model
- This approach can be used to standardize terminology in a specific field
- Ontologies can link multiple languages
- Foundation of the Semantic Web (not covered here)





Indexing of Documents

- An inverted index maps each keyword K_i to a set of documents S_i that contain the keyword
 - Documents identified by identifiers
- Inverted index may record
 - Keyword locations within document to allow proximity based ranking
 - Counts of number of occurrences of keyword to compute TF
- **and** operation: Finds documents that contain all of K_1 , K_2 , ..., K_n .
 - Intersection $S_1 \cap S_2 \cap \cap S_n$
- **or** operation: documents that contain at least one of $K_1, K_2, ..., K_n$
 - union, $S_1 \cap S_2 \cap \dots \cap S_n$,
- \blacksquare Each S_i is kept sorted to allow efficient intersection/union by merging
 - "not" can also be efficiently implemented by merging of sorted lists



Measuring Retrieval Effectiveness

- Information-retrieval systems save space by using index structures that support only approximate retrieval. May result in:
 - false negative (false drop) some relevant documents may not be retrieved.
 - false positive some irrelevant documents may be retrieved.
 - For many applications a good index should not permit any false drops, but may permit a few false positives.
- Relevant performance metrics:
 - precision what percentage of the retrieved documents are relevant to the query.
 - recall what percentage of the documents relevant to the query were retrieved.





Measuring Retrieval Effectiveness (Cont.)

- Recall vs. precision tradeoff:
 - Can increase recall by retrieving many documents (down to a low level of relevance ranking), but many irrelevant documents would be fetched, reducing precision
- Measures of retrieval effectiveness:
 - Recall as a function of number of documents fetched, or
 - Precision as a function of recall
 - Equivalently, as a function of number of documents fetched
 - E.g. "precision of 75% at recall of 50%, and 60% at a recall of 75%"
- Problem: which documents are actually relevant, and which are not





Web Search Engines

- Web crawlers are programs that locate and gather information on the Web
 - Recursively follow hyperlinks present in known documents, to find other documents
 - Starting from a seed set of documents
 - Fetched documents
 - Handed over to an indexing system
 - Can be discarded after indexing, or store as a cached copy
- Crawling the entire Web would take a very large amount of time
 - Search engines typically cover only a part of the Web, not all of it
 - Take months to perform a single crawl





Web Crawling (Cont.)

- Crawling is done by multiple processes on multiple machines, running in parallel
 - Set of links to be crawled stored in a database
 - New links found in crawled pages added to this set, to be crawled later
- Indexing process also runs on multiple machines
 - Creates a new copy of index instead of modifying old index
 - Old index is used to answer queries
 - After a crawl is "completed" new index becomes "old" index
- Multiple machines used to answer queries
 - Indices may be kept in memory
 - Queries may be routed to different machines for load balancing





Information Retrieval and Structured Data

- Information retrieval systems originally treated documents as a collection of words
- Information extraction systems infer structure from documents, e.g.:
 - Extraction of house attributes (size, address, number of bedrooms, etc.) from a text advertisement
 - Extraction of topic and people named from a new article
- Relations or XML structures used to store extracted data
 - System seeks connections among data to answer queries
 - Question answering systems





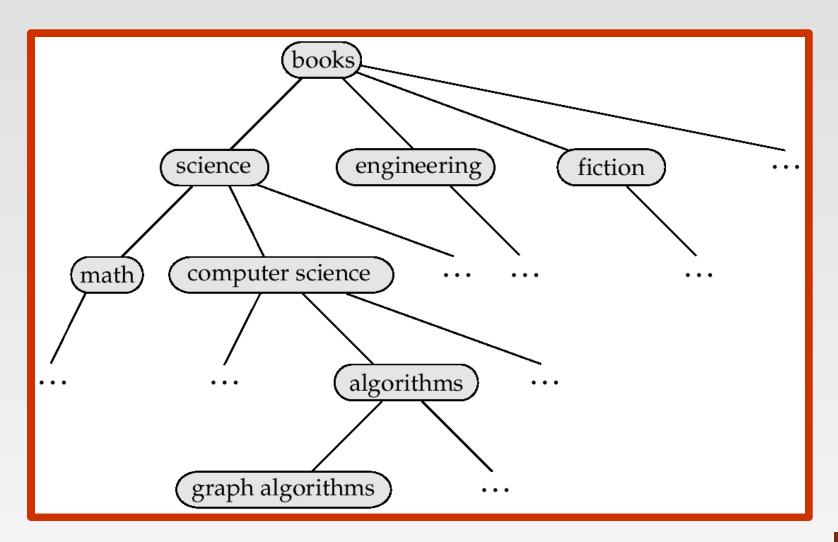
Directories

- Storing related documents together in a library facilitates browsing
 - users can see not only requested document but also related ones.
- Browsing is facilitated by classification system that organizes logically related documents together.
- Organization is hierarchical: classification hierarchy





A Classification Hierarchy For A Library System







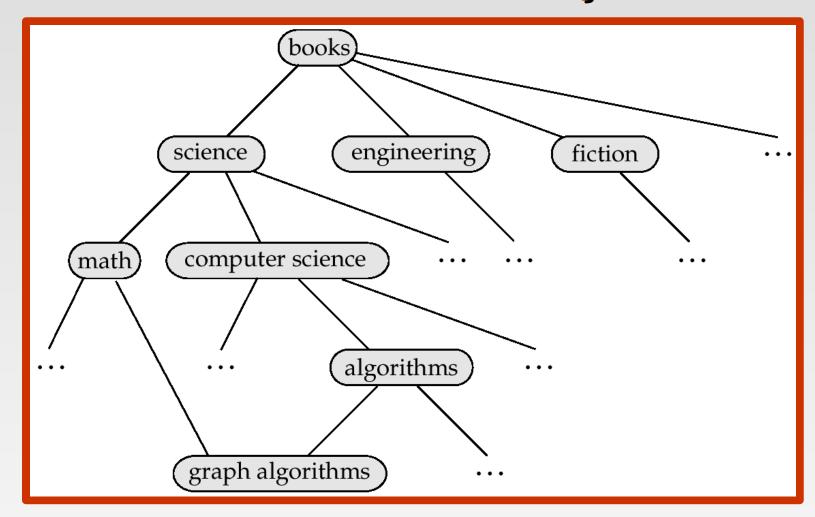
Classification DAG

- Documents can reside in multiple places in a hierarchy in an information retrieval system, since physical location is not important.
- Classification hierarchy is thus Directed Acyclic Graph (DAG)





A Classification DAG For A Library Information Retrieval System







Web Directories

- A Web directory is just a classification directory on Web pages
 - E.g. Yahoo! Directory, Open Directory project
 - Issues:
 - What should the directory hierarchy be?
 - Given a document, which nodes of the directory are categories relevant to the document
 - Often done manually
 - Classification of documents into a hierarchy may be done based on term similarity





End of Chapter

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