Lecture 1: Introduction and the Boolean Model Information Retrieval Computer Science Tripos Part II

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¹Based on slides from Simone Teufel and Ronan Cummins

1 Motivation

- Definition of "Information Retrieval"
- IR: beginnings to now

2 First Boolean Example

- Term–Document Incidence matrix
- The inverted index
- Processing Boolean Queries
- Practicalities of Boolean Search

Manning et al, 2008:

Information retrieval (IR) is finding material \ldots of an unstructured nature \ldots that satisfies an information need from within large collections \ldots .

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Document Collections





MS 3391 Library catalogue. Babylonia, 2000-1600 BC

Document Collections



IR in the 17th century: Samuel Pepys, the famous English diarist, subject-indexed his treasured 1000+ books library with key words.

Document Collections





Manning et al, 2008:

Information retrieval (IR) is finding material (usually documents) of an unstructured nature ... that satisfies an information need from within large collections (usually stored on computers).

- Document Collection: units we have built an IR system over.
- Usually documents
- But could be
 - memos
 - book chapters
 - paragraphs
 - scenes of a movie
 - turns in a conversation...
- Lots of them

Manning et al, 2008:

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Structured vs Unstructured Data

Unstructured data means that a formal, semantically overt, easy-for-computer structure is missing.

• In contrast to the rigidly structured data used in DB style searching (e.g. product inventories, personnel records)

Search Businesses	
Name / Type	
florists	
Location	
CB1	
Advanced Business Search	Search

SELECT *
FROM business_catalogue
WHERE category = 'florist'
AND city_zip = 'cb1'

- This does not mean that there is no structure in the data
 - Document structure (headings, paragraphs, lists...)
 - Explicit markup formatting (e.g. in HTML, XML...)
 - Linguistic structure (latent, hidden)

Manning et al, 2008:

Information retrieval (IR) is finding material (usually documents) of an unstructured nature (usually text) that satisfies an information need from within large collections (usually stored on computers).

- An information need is the topic about which the user desires to know more about.
- A query is what the user conveys to the computer in an attempt to communicate the information need.

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- Known-item search
- Precise information seeking search
- Open-ended search ("topical search")

Information scarcity vs. information abundance

- Information scarcity problem (or needle-in-haystack problem): hard to find rare information
 - Lord Byron's first words? 3 years old? Long sentence to the nurse in perfect English?

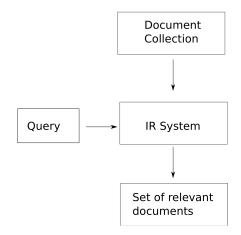
... when a servant had spilled an urn of hot coffee over his legs, he replied to the distressed inquiries of the lady of the house, 'Thank you, madam, the agony is somewhat abated.' [not Lord Byron, but Lord Macaulay]

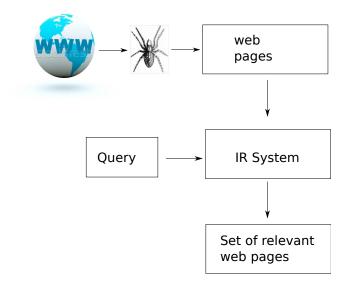
- Information abundance problem (for more clear-cut information needs): redundancy of obvious information
 - What is toxoplasmosis?

Manning et al, 2008:

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- A document is relevant if the user perceives that it contains information of value with respect to their personal information need.
- Are the retrieved documents
 - about the target subject
 - up-to-date?
 - from a trusted source?
 - satisfying the user's needs?
- How should we rank documents in terms of these factors?
- More on this in a lecture soon





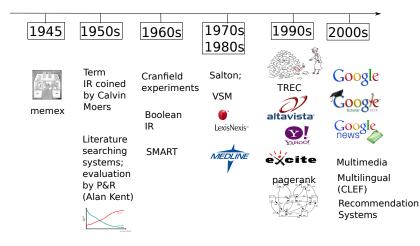
The effectiveness of an IR system (i.e., the quality of its search results) is determined by two key statistics about the system's returned results for a query:

- Precision: What fraction of the returned results are relevant to the information need?
- Recall: What fraction of the relevant documents in the collection were returned by the system?
- What is the best balance between the two?
 - Easy to get perfect recall: just retrieve everything
 - Easy to get good precision: retrieve only the most relevant

There is much more to say about this - lecture 6

- Web search (Google > bing)
 - Search ground are billions of documents on millions of computers
 - issues: spidering; efficient indexing and search; malicious manipulation to boost search engine rankings
 - Link analysis covered in Lecture 8
- Enterprise and institutional search (Publiced LexisNexis-)
 - e.g company's documentation, patents, research articles
 - often domain-specific
 - Centralised storage; dedicated machines for search.
 - Most prevalent IR evaluation scenario: US intelligence analyst's searches
- Personal information retrieval (email, pers. documents; 9)
 - e.g., Mac OS X Spotlight; Windows' Instant Search
 - Issues: different file types; maintenance-free, lightweight to run in background

A short history of IR



IR for non-textual media

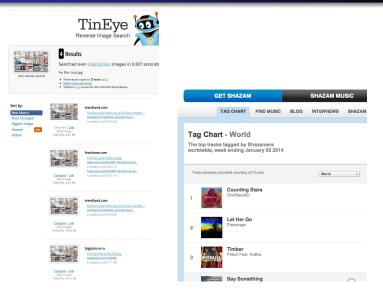








Similarity Searches



Motivation

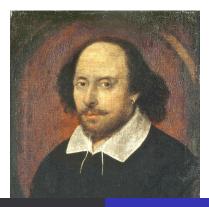
- Definition of "Information Retrieval"
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2 First Boolean Example

- Term–Document Incidence matrix
- The inverted index
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Boolean Retrieval Model

- In the Boolean retrieval model we can pose any query in the form of a Boolean expression of terms.
- i.e., one in which terms are combined with the operators AND, OR, and NOT.
- Model views each document as just a set of words.
- Example with Shakespeare's Collected works...



Brutus AND Caesar AND NOT Calpurnia

- Which plays of Shakespeare contain the words Brutus and Caesar, but not Calpurnia?
- Naive solution: linear scan through all text "grepping"
- In this case, works OK (Shakespeare's Collected works has less than 1M words).
- But in the general case, with much larger text colletions, we need to index.
- Indexing is an offline operation that collects data about which words occur in a text, so that at search time you only have to access the pre-compiled index.

Main idea: record for each document whether it contains each word out of all the different words Shakespeare used (about 32K).

	Antony and Cleopatra	Julius Caesar	The Tempest	Hamlet	Othello	Macbeth
Antony	1	1	0	0	0	1
Brutus	1	1	0	1	0	0
Caesar	1	1	0	1	1	1
Calpurnia	0	1	0	0	0	0
Cleopatra	1	0	0	0	0	0
mercy	1	0	1	1	1	1
worser	1	0	1	1	1	0

Matrix element (t, d) is 1 if the play in column d contains the word in row t, and 0 otherwise.

Query "Brutus AND Caesar AND NOT Calpurnia"

To answer the query, we take the vectors for *Brutus*, *Caesar* and *Calpurnia* (complement), and then do a bitwise AND:

	Antony and Cleopatra	Julius Caesar	The Tempest	Hamlet	Othello	Macbeth
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mercy	1	0	1	1	1	1
worser	1	0	1	1	1	0
AND	1	0	0	1	0	0

Bitwise AND returns two documents, "Antony and Cleopatra" and "Hamlet".

Antony and Cleopatra, Act III, Scene ii

Agrippa [Aside to Dominitus Enobarbus]: Why, Enobarbus, When Antony found Julius Caesar dead, He cried almost to roaring, and he wept When at Philippi he found Brutus slain.

Hamlet, Act III, Scene ii

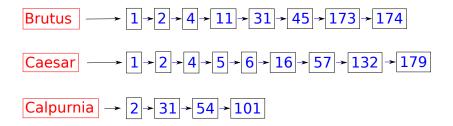
Lord Polonius: I did enact Julius Caesar: I was killed i' the Capitol; Brutus killed me.

- $\bullet~\mbox{Consider}~\mbox{N}{=}10^{6}$ documents, each ${\sim}1000$ words long
- 10^9 words at avg 6 bytes per word \Rightarrow 6GB
- \bullet Assume there are M=500,000 distinct terms in the collection
- $\bullet\,$ Size of incidence matrix is then 500,000 $\times 10^{6}\,$
- Half a trillion 0s and 1s

- Observation: the term-document matrix is very sparse.
- Contains no more than one billion 1s.
- Better representation: only record the things that do occur.
- Term-document matrix has other disadvantages, such as lack of support for more complex query operators (e.g., proximity search)
- We will move towards richer representations, beginning with the inverted index.

The inverted index consists of:

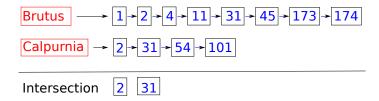
- a dictionary of terms (also: lexicon, vocabulary)
- and a postings list for each term, i.e., a list that records in which documents the term occurs (each item in the list is called a posting).



Our Boolean Query

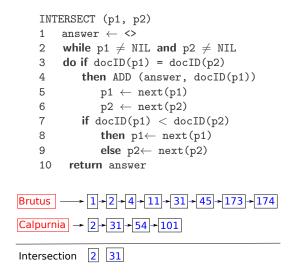
Brutus AND Calpurnia

Locate the postings lists of both query terms and intersect them.



Note: this only works if postings lists are sorted

Algorithm for intersection of two postings



Complexity of the Intersection Algorithm

- Bounded by worst-case length of postings lists
- Thus, formally, querying complexity is O(N), with N the number of documents in the document collection
- But in practice much, much better than linear scanning, which is asymptotically also O(N)

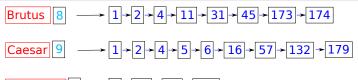
Query Optimisation: conjunctive terms

Organise order in which the postings lists are accessed so that least work needs to be done.

Brutus AND Caesar AND Calpurnia

Heuristic: process terms in order of increasing document frequency:

(Calpurnia AND Brutus) AND Caesar



Calpurnia $4 \rightarrow 2 \rightarrow 31 \rightarrow 54 \rightarrow 101$

(maddening OR crowd) AND (ignoble OR strife) AND (killed OR slain)

- Get the frequencies for all terms
- Estimate the size of each OR by the sum of the frequencies of its disjuncts (conservative)
- Process the query in increasing order of the size of each disjunctive term

Practical Boolean Search

- Provided by large commercial information providers 1960s-1990s
- Complex query language; complex and long queries
- Extended Boolean retrieval models with additional operators proximity operators
- Proximity operator: two terms must occur close together in a document (in terms of certain number of words, or within sentence or paragraph)
- Unordered results...

Commercial Boolean Searching Examples

- Westlaw : Largest commercial legal search service 500K subscribers
- Medical search
- Patent search
- Useful when expert queries are carefully defined and incrementally developed

On Google, the default interpretation of a query $[w_1 \; w_2 \; ... \; w_n]$ is $w_1 \; \text{AND} \; w_2 \; \text{AND} \; ... \; \text{AND} \; w_n$

- Cases where you get hits which don't contain one of the w_i:
 - Page contains variant of w_i (morphology, misspelling, synonym)
 - long query (n is large)
 - Boolean expression generates very few hits
 - w_i was in the anchor text
- Google also *ranks* the result set
 - Simple Boolean Retrieval returns matching documents in no particular order.
 - Google (and most well-designed Boolean engines) rank hits according to some estimator of relevance

• Manning, Raghavan, Schütze: Introduction to Information Retrieval (MRS), chapter 1