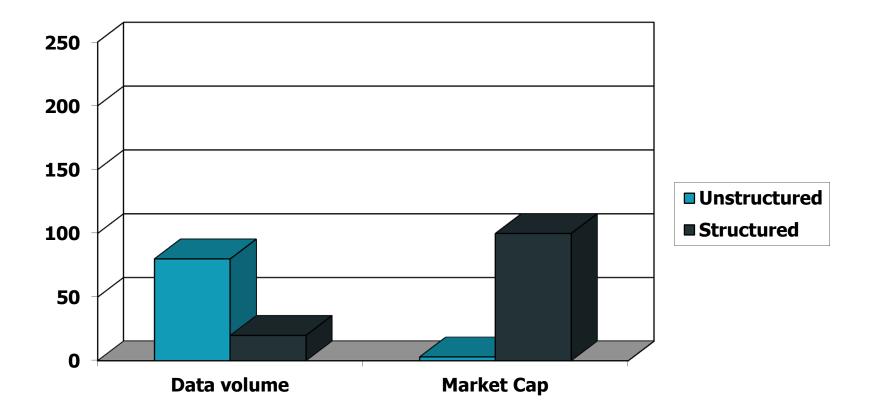
Introduction to Information Retrieval

Introducing Information Retrieval and Web Search

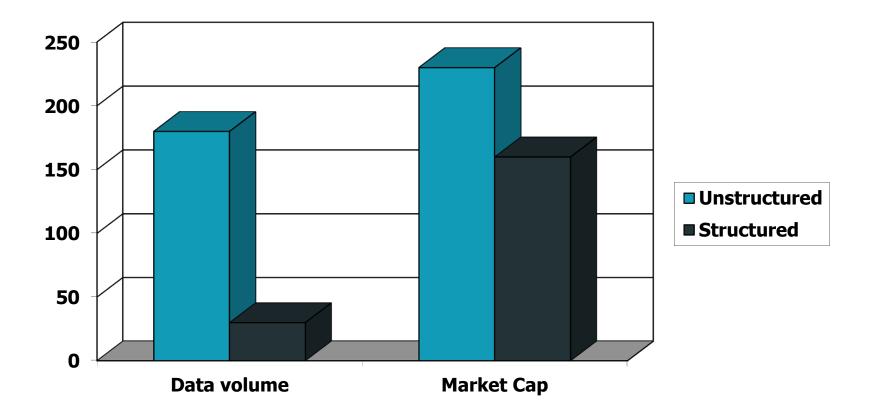
Information Retrieval

- 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).
 - These days we frequently think first of web search, but there are many other cases:
 - E-mail search
 - Searching your laptop
 - Corporate knowledge bases
 - Legal information retrieval

Unstructured (text) vs. structured (database) data in the mid-nineties



Unstructured (text) vs. structured (database) data today

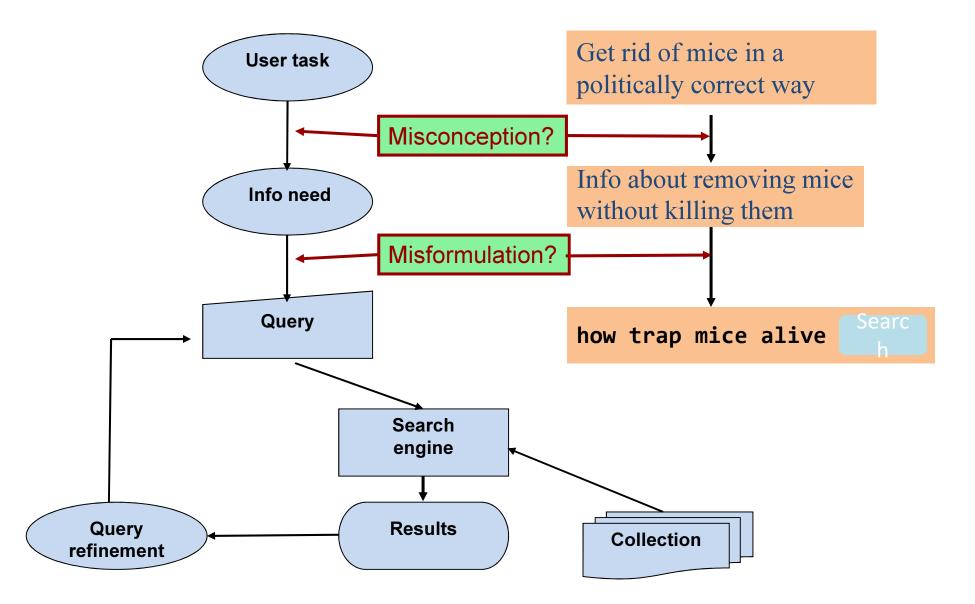


Basic assumptions of Information Retrieval

- Collection: A set of documents

 Assume it is a static collection for the moment
- Goal: Retrieve documents with information that is relevant to the user's information need and helps the user complete a task

The classic search model



How good are the retrieved docs?

- Precision : Fraction of retrieved docs that are relevant to the user's information need
- *Recall* : Fraction of relevant docs in collection that are retrieved
 - More precise definitions and measurements to follow later

Introduction to Information Retrieval

Term-document incidence matrices

Unstructured data in 1620

- Which plays of Shakespeare contain the words Brutus AND Caesar but NOT Calpurnia?
- One could grep all of Shakespeare's plays for Brutus and Caesar, then strip out lines containing Calpurnia?
- Why is that not the answer?
 - Slow (for large corpora)
 - <u>NOT</u> Calpurnia is non-trivial
 - Other operations (e.g., find the word *Romans* near *countrymen*) not feasible
 - Ranked retrieval (best documents to return)
 - Later lectures

Term-document incidence matrices

	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
Brutus Calpur	AND Caesar E nia	BUT NOT		l if pla vord, (-	

Incidence vectors

- So we have a 0/1 vector for each term.
- To answer query: take the vectors for *Brutus, Caesar* and *Calpurnia* (complemented) →
 bitwise AND.
 - 110100 AND
 - 110111 *ΑΝΓ*
 - 101111 =
 - 100100

	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

Answers to query

Antony and Cleopatra, Act III, Scene ii

Agrippa [Aside to DOMITIUS 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.



Bigger collections

- Consider N = 1 million documents, each with about 1000 words.
- Avg 6 bytes/word including spaces/punctuation
 - 6GB of data in the documents.
- Say there are *M* = 500K *distinct* terms among these.

Why?

Can't build the matrix

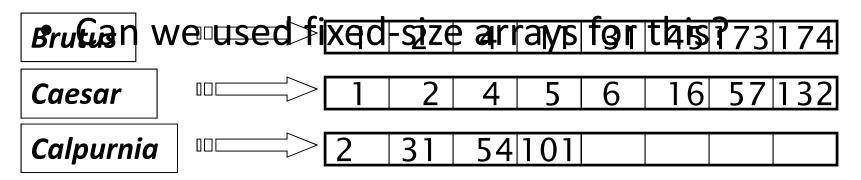
- 500K x 1M matrix has half-a-trillion 0's and 1's.
- But it has no more than one billion 1's.
 matrix is extremely sparse.
- What's a better representation?
 We only record the 1 positions.

Introduction to Information Retrieval

The Inverted Index The key data structure underlying modern IR

Inverted index

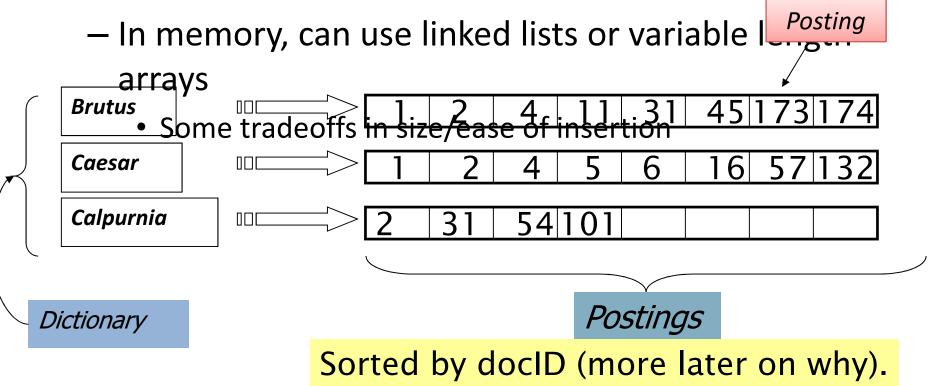
- For each term *t*, we must store a list of all documents that contain *t*.
 - Identify each doc by a **docID**, a document serial number



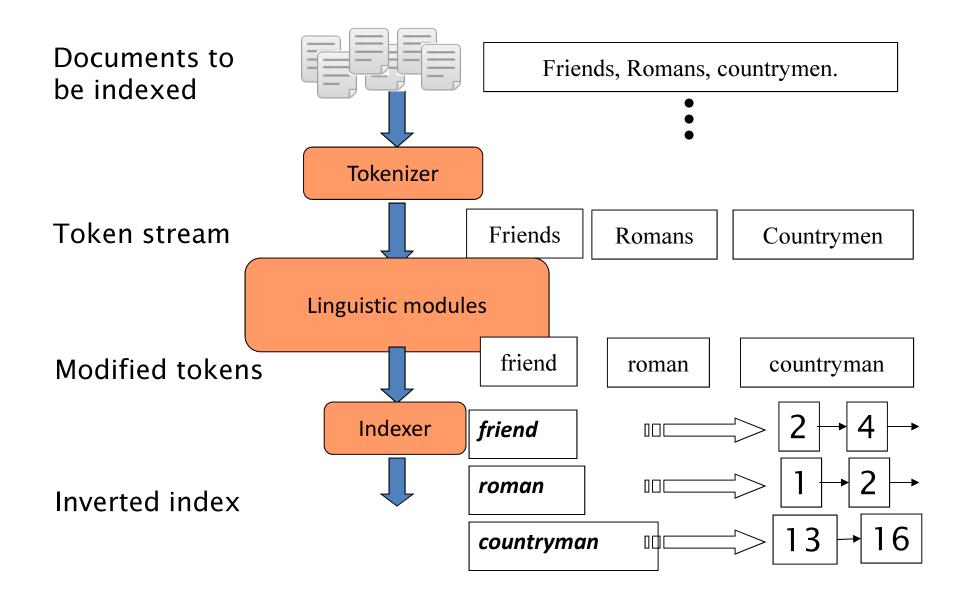
What happens if the word *Caesar* is added to document 14?

Inverted index

- We need variable-size postings lists
 - On disk, a continuous run of postings is normal and best



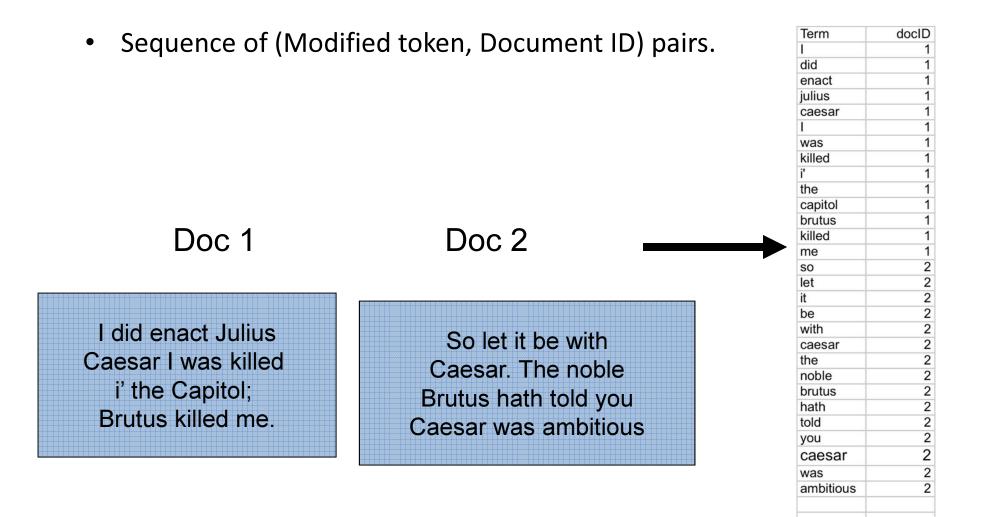
Inverted index construction



Initial stages of text processing

- Tokenization
 - Cut character sequence into word tokens
 - Deal with "John's", a state-of-the-art solution
- Normalization
 - Map text and query term to same form
 - You want **U.S.A.** and **USA** to match
- Stemming
 - We may wish different forms of a root to match
 - authorize, authorization
- Stop words
 - We may omit very common words (or not)
 - the, a, to, of

Indexer steps: Token sequence



Indexer steps: Sort

• Sort by terms

And then docID



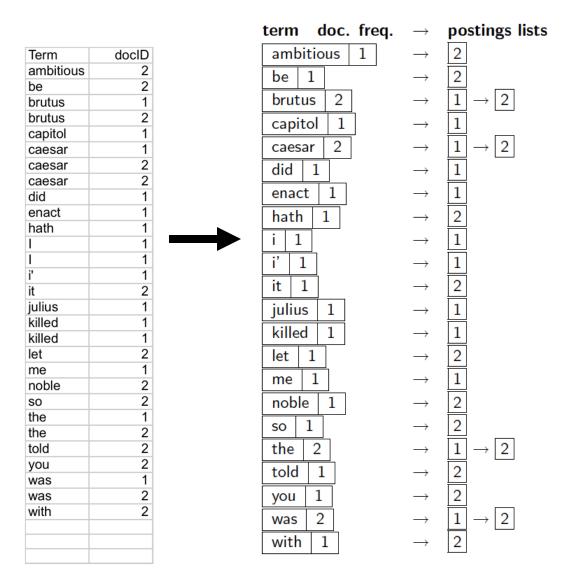
Term	docID		Te
I	1		an
did	1		be
enact	1		br
julius	1		br
caesar	1		ca
I	1		ca
was	1		ca
killed	1		ca
i'	1		dia
the	1		en
capitol	1		ha
brutus	1		1
killed	1		I
me	1	$\overline{}$	i'
so	2		it
let	2		jul
it	2		kil
be	2		kil
with	2		let
caesar	2		m
the	2		nc
noble	2		so
brutus	2		the
hath	2		the
told	2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2		tol
you	2		yo
caesar	2		Wa
was	2		Wa
ambitious	2		wi

Term	docID
ambitious	2 2 1 2 1 1 2 2 2 2 1
be	2
brutus	1
brutus	2
capitol	1
caesar	1
caesar	2
caesar	2
did	1
enact	1
hath	1
	1
1	1
i'	1
it	2
julius	1
killed	1
killed	1
let	2
me	1
noble	2
SO	2
the	1
the	2
told	2
you	2
was	1
was	1 1 2 1 1 2 1 2 2 1 2 2 2 2 2 1 2 2 2 2
with	2

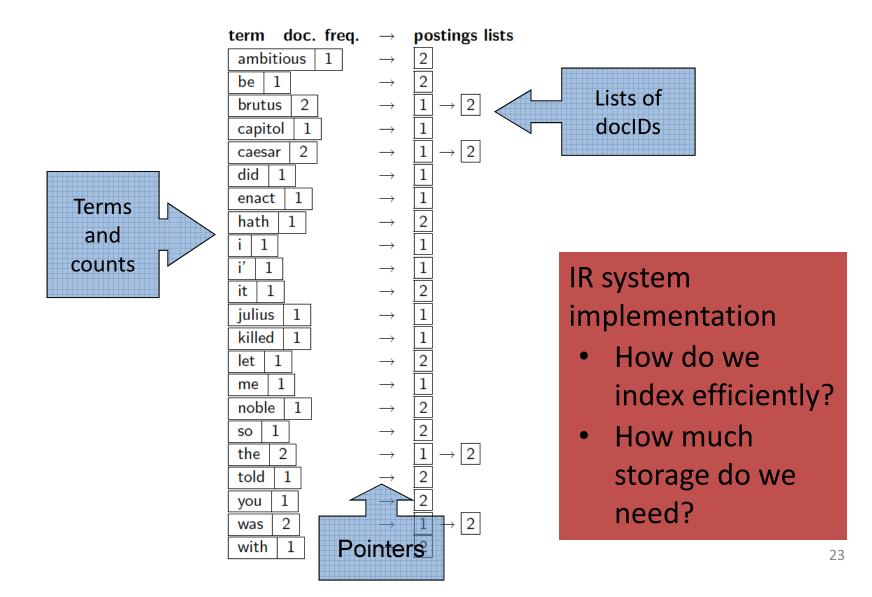
Indexer steps: Dictionary & Postings

- Multiple term entries in a single document are merged.
- Split into Dictionary and Postings
- Doc. frequency information is added.

Why frequency? Will discuss later.



Where do we pay in storage?



Introduction to Information Retrieval

Query processing with an inverted index

The index we just built

• How do we process a query?



- Later - what kinds of queries can we process?

Query processing: AND

• Consider processing the query:

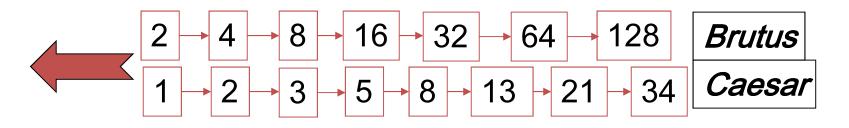
Brutus AND Caesar

- Locate **Brutus** in the Dictionary;
 - Retrieve its postings.
- Locate *Caesar* in the Dictionary;
 - Retrieve its postings.

 "Merge" the two postings (intersect the document 2 32 8 16 128 sets): 64 Brutus 4 3 Caesar 2 → 5 8 → 13 21 34

The merge

 Walk through the two postings simultaneously, in time linear in the total number of postings entries



If the list lengths are x and y, the merge takes O(x+y) operations. Crucial: postings sorted by docID.

Intersecting two postings lists (a "merge" algorithm)

INTERSECT (p_1, p_2)

- 1 answer $\leftarrow \langle \rangle$
- 2 while $p_1 \neq \text{NIL}$ and $p_2 \neq \text{NIL}$
- 3 **do if** $docID(p_1) = docID(p_2)$
- 4 then $ADD(answer, doc ID(p_1))$
- 5 $p_1 \leftarrow next(p_1)$ 6 $p_2 \leftarrow next(p_2)$
 - $p_2 \leftarrow next(p_2)$
- 7 else if $doclD(p_1) < doclD(p_2)$ 8 then $p_1 \leftarrow next(p_1)$ 9 else $p_2 \leftarrow next(p_2)$

10 return answer

Introduction to Information Retrieval

The Boolean Retrieval Model & Extended Boolean Models

Boolean queries: Exact match

- The Boolean retrieval model is being able to ask a query that is a Boolean expression:
 - Boolean Queries are queries using AND, OR and NOT to join query terms
 - Views each document as a <u>set</u> of words
 - Is precise: document matches condition or not.
 - Perhaps the simplest model to build an IR system on
- Primary commercial retrieval tool for 3 decades.
- Many search systems you still use are Boolean:
 - Email, library catalog, Mac OS X Spotlight

Example: WestLaw http://www.westlaw.com/

- Largest commercial (paying subscribers) legal search service (started 1975; ranking added 1992; new federated search added 2010)
- Tens of terabytes of data; ~700,000 users
- Majority of users *still* use boolean queries
- Example query:
 - What is the statute of limitations in cases involving the federal tort claims act?
 - LIMIT! /3 STATUTE ACTION /S FEDERAL /2 TORT /3 CLAIM
 - /3 = within 3 words, /S = in same sentence

Example: WestLaw http://www.westlaw.com/

- Another example query:
 - Requirements for disabled people to be able to access a workplace
 - disabl! /p access! /s work-site work-place (employment /3 place
- Note that SPACE is disjunction, not conjunction!
- Long, precise queries; proximity operators; incrementally developed; not like web search
- Many professional searchers still like Boolean search

You know exactly what you are getting

• But that doesn't mean it actually works better....

Boolean queries: More general merges

- Exercise: Adapt the merge for the queries:
 Brutus AND NOT Caesar
 Brutus OR NOT Caesar
- Can we still run through the merge in time
 O(x+y)? What can we achieve?

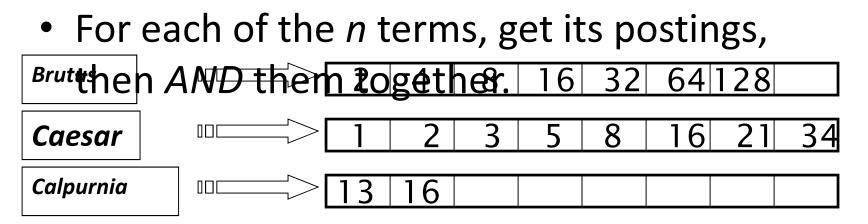
Merging

What about an arbitrary Boolean formula? (Brutus OR Caesar) AND NOT (Antony OR Cleopatra)

- Can we always merge in "linear" time?
 Linear in what?
- Can we do better?

Query optimization

- What is the best order for query processing?
- Consider a query that is an AND of *n* terms.

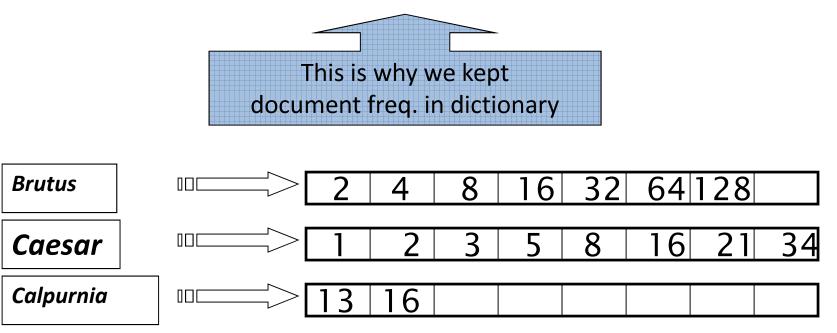


Query: Brutus AND Calpurnia AND Caesar

Query optimization example

• <u>Process in order of increasing freq</u>:

- start with smallest set, then keep cutting further.



Execute the query as (*Calpurnia* AND *Brutus)* AND *Caesar*.

More general optimization

- e.g., (madding OR crowd) AND (ignoble OR strife)
- Get doc. freq.'s for all terms.
- Estimate the size of each *OR* by the sum of its doc. freq.'s (conservative).
- Process in increasing order of OR sizes.

Exercise

 Recommend a query processing order for

(tangerine OR trees) AND (marmalade OR skies) AND (kaleidoscope OR eyes)

• Which two terms should we process first?

Term	Freq
eyes	213312
kaleidoscope	87009
marmalade	107913
skies	271658
tangerine	46653
trees	316812

Query processing exercises

- Exercise: If the query is *friends* AND *romans* AND (NOT countrymen), how could we use the freq of countrymen?
- Exercise: Extend the merge to an arbitrary Boolean query. Can we always guarantee execution in time linear in the total postings size?
- Hint: Begin with the case of a Boolean *formula* query: in this, each query term appears only once in the query.

Exercise

- Try the search feature at <u>http://www.rhymezone.com/shakespeare/</u>
- Write down five search features you think it could do better

Introduction to Information Retrieval

Phrase queries and positional indexes

Phrase queries

- We want to be able to answer queries such as *"stanford university"* – as a phrase
- Thus the sentence *"I went to university at Stanford"* is not a match.
 - The concept of phrase queries has proven easily understood by users; one of the few "advanced search" ideas that works
 - Many more queries are *implicit phrase queries*
- For this, it no longer suffices to store only <term : docs> entries

A first attempt: Biword indexes

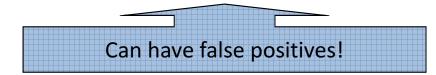
- Index every consecutive pair of terms in the text as a phrase
- For example the text "Friends, Romans, Countrymen" would generate the biwords
 - friends romans
 - romans countrymen
- Each of these biwords is now a dictionary term
- Two-word phrase query-processing is now immediate.

Longer phrase queries

- Longer phrases can be processed by breaking them down
- stanford university palo alto can be broken into the Boolean query on biwords:

stanford university AND university palo AND palo alto

Without the docs, we cannot verify that the docs matching the above Boolean query do contain the phrase.



Issues for biword indexes

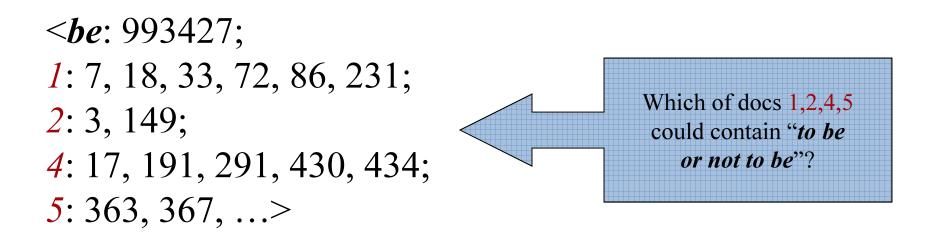
- False positives, as noted before
- Index blowup due to bigger dictionary
 - Infeasible for more than biwords, big even for them
- Biword indexes are not the standard solution (for all biwords) but can be part of a compound strategy

Solution 2: Positional indexes

• In the postings, store, for each *term* the position(s) in which tokens of it appear:

<term, number of docs containing term; doc1: position1, position2 ... ; doc2: position1, position2 ... ; etc.>

Positional index example



- For phrase queries, we use a merge algorithm recursively at the document level
- But we now need to deal with more than just equality

Processing a phrase query

- Extract inverted index entries for each distinct term: *to, be, or, not.*
- Merge their *doc:position* lists to enumerate all positions with "*to be or not to be*".
 - **to**:
 - 2:1,17,74,222,551; **4:8,16,190,429,433;** 7:13,23,191; ...
 - **be**:
 - 1:17,19; **4:17,191,291,430,434;** 5:14,19,101; ...
- Same general method for proximity searches

Proximity queries

- LIMIT! /3 STATUTE /3 FEDERAL /2 TORT
 Again, here, /k means "within k words of".
- Clearly, positional indexes can be used for such queries; biword indexes cannot.
- Exercise: Adapt the linear merge of postings to handle proximity queries. Can you make it work for any value of k?
 - This is a little tricky to do correctly and efficiently
 - See Figure 2.12 of *IIR*

Positional index size

- A positional index expands postings storage substantially
 - Even though indices can be compressed
- Nevertheless, a positional index is now standardly used because of the power and usefulness of phrase and proximity queries ... whether used explicitly or implicitly in a ranking retrieval system.

Positional index size

- Need an entry for each occurrence, not just once per document
- Index size depends on average document size^{Why?}
 - Average web page has <1000 terms
 - SEC filings, books, even some epic poems ... easily 100,000 terms
- Consider a term with frequency 0.1%

Document size	Postings	Positional postings
1000	1	1
100,000	1	100

Rules of thumb

- A positional index is 2–4 as large as a nonpositional index
- Positional index size 35–50% of volume of original text
 - Caveat: all of this holds for "English-like" languages

Combination schemes

- These two approaches can be profitably combined
 - For particular phrases (*"Michael Jackson", "Britney Spears"*) it is inefficient to keep on merging positional postings lists
 - Even more so for phrases like "The Who"
- Williams et al. (2004) evaluate a more sophisticated mixed indexing scheme
 - A typical web query mixture was executed in ¼ of the time of using just a positional index
 - It required 26% more space than having a positional index alone

Introduction to Information Retrieval

Structured vs. Unstructured Data

IR vs. databases: Structured vs unstructured data

 Structured data tends to refer to information in "tables"

Employee	Manager	Salary
Smith	Jones	50000
Chang	Smith	60000
lvy	Smith	50000

Typically allows numerical range and exact match (for text) queries, e.g., Salary < 60000 AND Manager = Smith.

Unstructured data

- Typically refers to free text
- Allows
 - Keyword queries including operators
 - More sophisticated "concept" queries e.g.,
 - find all web pages dealing with *drug abuse*
- Classic model for searching text documents

Semi-structured data

- In fact almost no data is "unstructured"
- E.g., this slide has distinctly identified zones such as the *Title* and *Bullets*
 - ... to say nothing of linguistic structure
- Facilitates "semi-structured" search such as
 - Title contains data AND Bullets contain search
- Or even
 - *Title* is about <u>Object Oriented Programming</u> AND Author something like <u>stro*rup</u>
 - where * is the wild-card operator