

Basic inverted indexes: Structure – Dictionary and Postings

- Key steps in construction sorting
- Boolean query processing
 - Simple optimization
 - Linear time merging
- Overview of course topics

Plan for this lecture

- Finish basic indexing
 - Tokenization
 - What terms do we put in the index?
- Query processing speedups
- Proximity/phrase queries





Tokenization

- Input: "Friends, Romans and Countrymen"
- <u>Output</u>: Tokens
 - Friends
 - Romans
 - Countrymen
- Each such token is now a candidate for an index entry, after <u>further processing</u>
 - Described below
- But what are valid tokens to emit?

Parsing a document

- What format is it in?
- pdf/word/excel/html?
- What language is it in?
- What character set is in use?

Each of these is a classification problem, which we will study later in the course.

But there are complications ...

Format/language stripping

- Documents being indexed can include docs from many different languages
 - A single index may have to contain terms of several languages.
- Sometimes a document or its components can contain multiple languages/formats
 - French email with a Portuguese pdf attachment.
- What is a unit document?
 - An email?
 - With attachments?
 - An email with a zip containing documents?

Tokenization

- Issues in tokenization:
 - Finland's capital → Finland? Finlands? Finland's?
 - *Hewlett-Packard* → *Hewlett* and *Packard* as two tokens?
 - San Francisco: one token or two? How do you decide it is one token?

Language issues

- Accents: résumé vs. resume.
- L'ensemble → one token or two?
 L?L'?Le?
- How are your users like to write their queries for these words?

Tokenization: language issues

- Chinese and Japanese have no spaces between words:
 - Not always guaranteed a unique tokenization
- Further complicated in Japanese, with multiple alphabets intermingled
 - Dates/amounts in multiple formats
- フォーチュン500社は情報不足のため時間あた\$500K(約6,000万円)

Katakana Hiragana Kanji "Romaji"

End-user can express query entirely in (say) Hiragana!

Normalization In "right-to-left languages" like Hebrew and Arabic: you can have "left-to-right" text interspersed (e.g., for dollar amounts). Need to "normalize" indexed text as well as query terms into the same form *Tf30 ff vs. T/30* Character-level alphabet detection and conversion Tokenization not separable from this. Sometimes ambiguous: Is this German "mit"?

Punctuation

- **Ne'er**: use language-specific, handcrafted "locale" to normalize.
 - Which language?
 - Most common: detect/apply language at a predetermined granularity: doc/paragraph.
- State-of-the-art: break up hyphenated sequence. Phrase index?
- U.S.A. vs. USA use locale.
- a.out

Numbers

- 3/12/91
- Mar. 12, 1991
- 55 B.C.
- B-52
- My PGP key is 324a3df234cb23e
- **100.2.86.144**
 - Generally, don't index as text.
 - Will often index "meta-data" separately
 - Creation date, format, etc.

Case folding

- Reduce all letters to lower case
 - exception: upper case (in mid-sentence?)
 - e.g., General Motors
 - Fed vs. fed
 - SAIL vs. sail

Thesauri and soundex

- Handle synonyms and homonyms
 - Hand-constructed equivalence classes
 - e.g., car = automobile
 - 🛚 your 🎝 you're
- Index such equivalences
 - When the document contains *automobile*, index it under *car* as well (usually, also vice-versa)
- Or expand query?
 - When the query contains *automobile*, look under *car* as well

Soundex

- Class of heuristics to expand a query into phonetic equivalents
 - Language specific mainly for names
 - E.g., chebyshev → tchebycheff
- More on this later ...

Lemmatization

- Reduce inflectional/variant forms to base form
- ∎ E.g.,
 - am, are, is \rightarrow be
 - car, cars, car's, cars' \rightarrow car
- the boy's cars are different colors → the boy car be different color





Porter's algorithm

- Commonest algorithm for stemming English
- Conventions + 5 phases of reductions
 - phases applied sequentially
 - each phase consists of a set of commands
 - sample convention: Of the rules in a compound command, select the one that applies to the longest suffix.

Typical rules in Porter

- sses → ss
- ies \rightarrow i
- ational \rightarrow ate
- tional \rightarrow tion

Other stemmers

- Other stemmers exist, e.g., Lovins stemmer http://www.comp.lancs.ac.uk/computing/research/stemming/general/lovins.htm
- Single-pass, longest suffix removal (about 250 rules)
- Motivated by Linguistics as well as IR
- Full morphological analysis modest benefits for retrieval

Language-specificity

- Many of the above features embody transformations that are
 - Language-specific and
 - Often, application-specific
- These are "plug-in" addenda to the indexing process
- Both open source and commercial plug-ins available for handling these











Placing skips

- Simple heuristic: for postings of length *L*, use √*L* evenly-spaced skip pointers.
- This ignores the distribution of query terms.
- Easy if the index is relatively static; harder if *L* keeps changing because of updates.



Phrase queries

- Want to answer queries such as stanford university – as a phrase
- Thus the sentence "I went to university at Stanford" is not a match.
- 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 are processed as we did with wild-cards:
- *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.

Can have false positives!

Extended biwords

- Parse the indexed text and perform part-ofspeech-tagging (POST).
- Bucket the terms into (say) Nouns (N) and articles/prepositions (X).
- Now deem any string of terms of the form NX*N to be an <u>extended biword</u>.
 - Each such extended biword is now made a term in the dictionary.
- Example:

catcher in the rye
 N X X N

Query processing

- Given a query, parse it into N's and X's
 Segment query into enhanced biwords
 - Look up index
 - LOOK up inc
- Issues
 - Parsing longer queries into conjunctions
 - E.g., the query tangerine trees and marmalade skies is parsed into
 - tangerine trees AND trees and marmalade AND marmalade skies

Other issues

- False positives, as noted before
- Index blowup due to bigger dictionary

Solution 2: Positional indexes

Store, for each *term*, entries of the form:
 <number of docs containing *term*;
 doc1: position1, position2 ...;
 doc2: position1, position2 ...;
 etc.>





Proximity queries

- LIMIT! /3 STATUTE /3 FEDERAL /2 TORT 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*?

Positional index size

- Can compress position values/offsets as we did with docs in the last lecture
- Nevertheless, this expands postings storage substantially

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

- Positional index size factor of 2-4 over nonpositional index
- Positional index size 35-50% of volume of original text
- Caveat: all of this holds for "English-like" languages

Resources for today's lecture

- MG 3.6, 4.3; MIR 7.2
- Porter's stemmer:
- http://www.sims.berkeley.edu/~hearst/irbook/porter.html
- H.E. Williams, J. Zobel, and D. Bahle, "Fast Phrase Querying with Combined Indexes", ACM Transactions on Information Systems.

http://www.seg.rmit.edu.au/research/research.php?author=4