# Introduction to Information Retrieval

Introducing Information Retrieval and Web Search

### Perkenalan

- Mata kuliah: Topik dalam Sistem Temu Kembali Informasi
- Semester 3 Program Studi Magister Teknik Informatika ITS
- Dosen: Dr. Agus Zainal Arifin
- Perkuliahan
  - Lokasi ruang kuliah: Gedung Rektorat lt. 3
  - Hari Senin 12:30-15:00
  - Sejak 28 Agustus 2017 selama 16 minggu
  - Komunikasi via Facebook dan Grup Whatsapp TD STKI
  - Lab dilaksanakan secara mandiri, sesuai kebutuhan
- Penilaian: Tes Tulis (20 %), Project (30 %), Makalah Paper (30 %), Presentasi Akhir (20 %). Apabila makalah diterima oleh jurnal yang bereputasi, maka penilaian diganti nilai A.

## Capaian Pembelajaran

#### Capaian Pembelajaran Prodi yang didukung:

- 1. Bertaqwa kepada Tuhan Yang Maha Esa dan mampu menunjukkan sikap religius; Sikap 1.
- 2. Menjunjung tinggi nilai kemanusiaan dalam menjalankan tugas berdasarkan agama, moral dan etika; Sikap 2.
- 3. Berkontribusi dalam peningkatan mutu kehidupan bermasyarakat, berbangsa, bernegara, dan peradaban berdasarkan Pancasila; Sikap 3.
- 4. Berperan sebagai warga negara yang bangga dan cinta tanah air, memiliki nasionalisme serta rasa tanggungjawab pada negara dan bangsa; Sikap 4.
- 5. Mampu menerapkan pemikiran logis, kritis, sistematis, dan inovatif dalam konteks pengembangan atau implementasi ilmu pengetahuan dan teknologi yang memperhatikan dan menerapkan nilai humaniora yang sesuai dengan bidang keahliannya; Keterampilan Umum 1.
- 6. Mampu menunjukkan kinerja mandiri, bermutu, dan terukur; Keterampilan Umum 2.

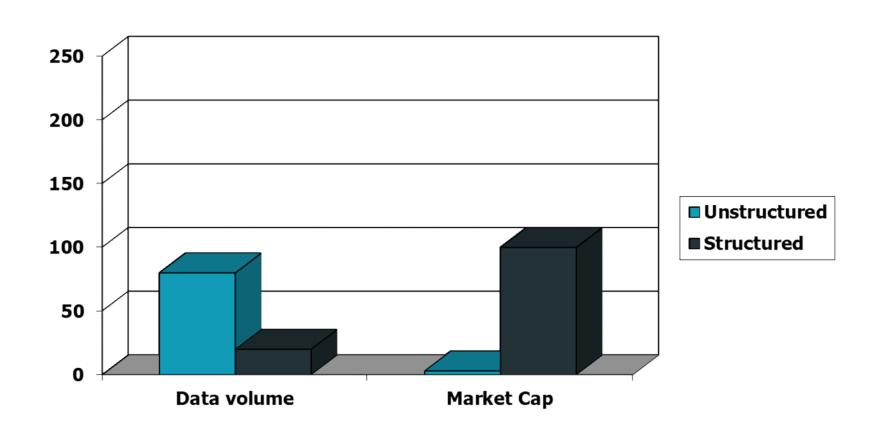
#### Capaian Pembelajaran Kuliah:

- S1: Mahasiswa mampu mengembangkan sistem yang dapat menemukan dokumen yang relevan dengan kebutuhan pengguna dengan menerapkan prinsip-prinsip dan algoritma yang efektif dalam pemilihan dokumen.
- 2. S2: Mahasiswa mampu mengembangkan prinsip dan algoritma yang efektif untuk memecahkan permasalahan dalam menemukan dokumen yang relevan dengan kebutuhan pengguna, hingga menghasilkan karya inovatif dan teruji.

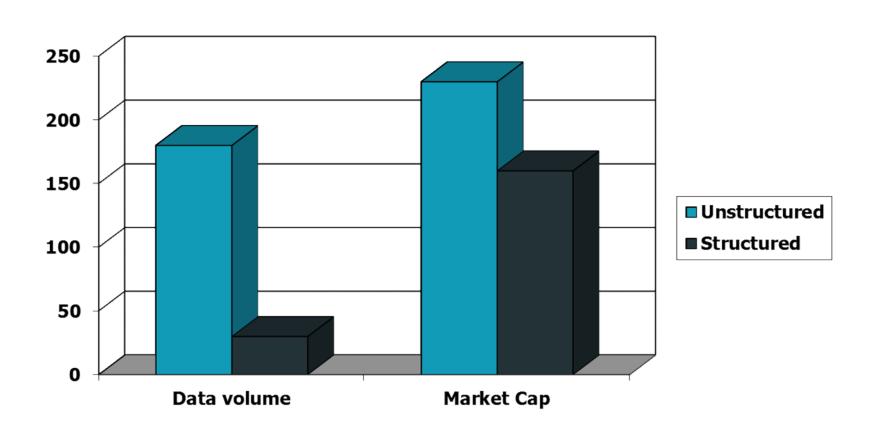
### Information Retrieval

- 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

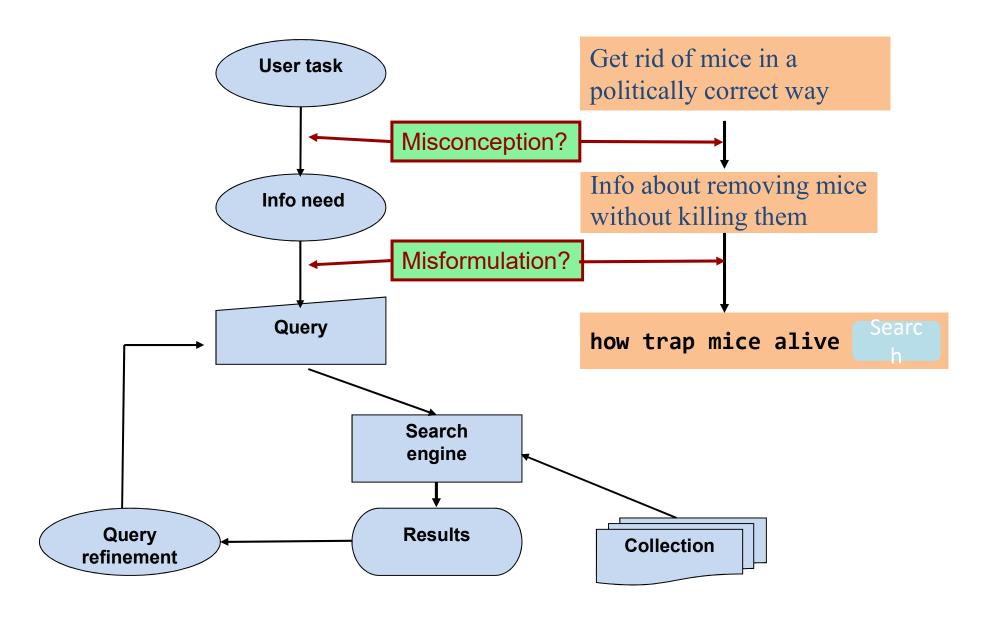


# Basic assumptions of Information Retrieval

- 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

## Kutipan Kalimat Shakespeare

- Apalah arti sebuah nama?
  - What's in a name?
- Jika musik adalah makanan bagi cinta, mainkanlah.
  - If music be the food of love, play on.
- Ketika seorang ayah memberi kepada anaknya, keduanya tertawa; ketika anak yang memberikan kepada ayahnya, keduanya menangis.
  - When a father gives to his son, both laugh; when a son gives to his father, both cry.

### 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)
  - NOT 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

|           | <b>Antony and Cleopatra</b> | <b>Julius Caesar</b> | 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       |

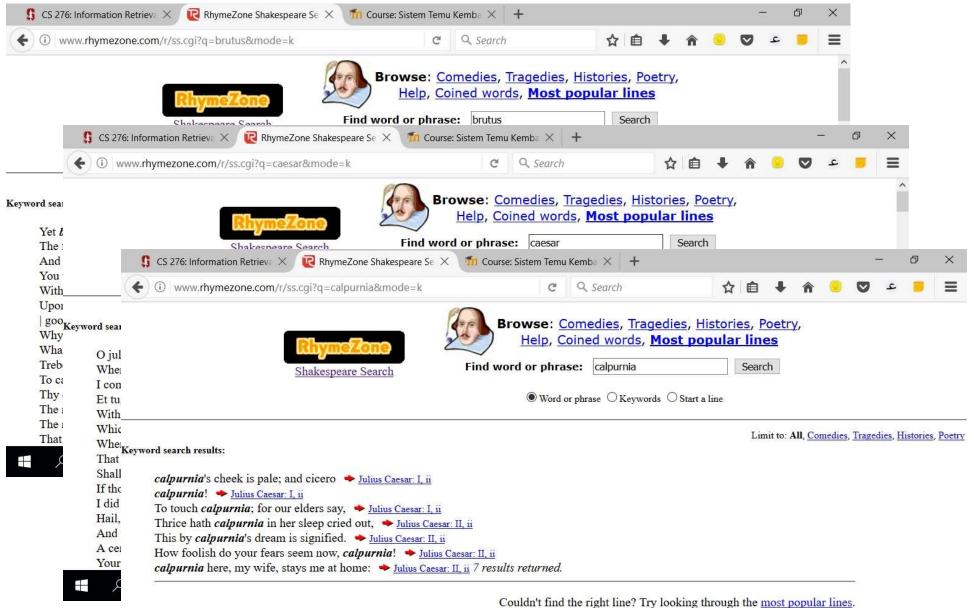
Brutus AND Caesar BUT NOT Calpurnia

1 if play contains word, 0 otherwise

### 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 *AND* 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.

#### Sec. 1.1

## 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.

### Can't build the matrix

- 500K x 1M matrix has half-a-trillion 0's and 1's.
  - $-5x10^2x10^3x10^6=5x10^{11}=0.5x10^{12}=$  setengah Trilyun
- But it has no more than one billion 1's.



- 1000 words x 1M documents = 1 Miliar words
- matrix is extremely sparse.
- What's a better representation?
  - We only record the 1 positions.

# Introduction to Information Retrieval

The Inverted Index

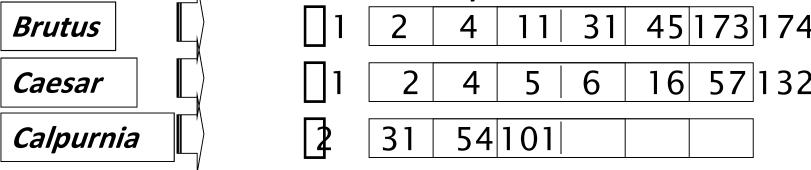
The key data structure underlying

modern IR

#### Sec. 1.2

### Inverted index

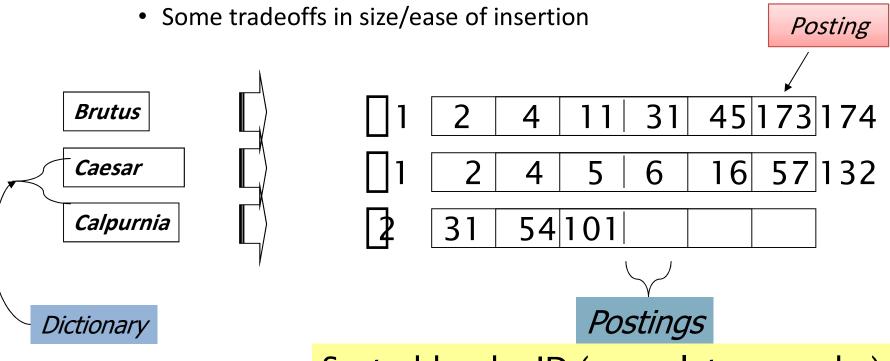
- For each term t, we must store a list of all documents that contain t.
  - Identify each doc by a docID, a document serial number
- Can we used fixed-size arrays for this?



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

## Inverted index

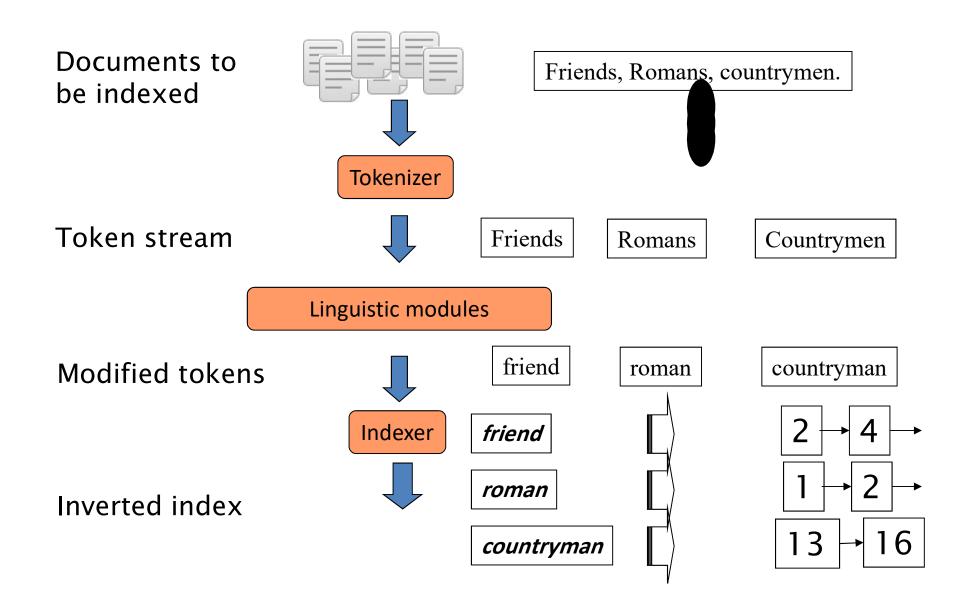
- We need variable-size postings lists
  - On disk, a continuous run of postings is normal and best
  - In memory, can use linked lists or variable length arrays



Sorted by docID (more later on why).

#### Sec. 1.2

### 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

#### 500. 1.2

# Indexer steps: Token sequence

• Sequence of (Modified token, Document ID) pairs.

Doc 1

I did enact Julius Caesar I was killed i' the Capitol; Brutus killed me. Doc 2

So let it be with Caesar. The noble Brutus hath told you Caesar was ambitious

| Term      | docID  |
|-----------|--|
| 1         | 1  |
| did       | 1  |
| enact     | 1  |
| julius    | 1  |
| caesar    | 1  |
| 1         | 1  |
| was       | 1  |
| killed    | 1  |
| i'        | 1  |
| the       | 1  |
| capitol   | 1  |
| brutus    | 1  |
| killed    | 1  |
| me        | 1  |
| so        | 2  |
| let       | 2  |
| it        | 2  |
| be        | 2  |
| with      | 2  |
| caesar    | 2  |
| the       | 2  |
| noble     | 2  |
| brutus    | 2  |
| hath      | 2  |
| told      | 2  |
| you       | 2  |
| caesar    | 1<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2 |
| was       | 2  |
| ambitious | 2  |
|           |  |
|           |  |
|           |  |

#### Sec. 1.2

# Indexer steps: Sort

- Sort by terms
  - And then docID

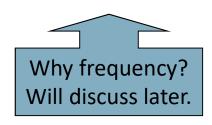


| Term      | docID   |
|-----------|---|
| 1         | 1   |
| did       | 1   |
| enact     | 1   |
| julius    | 1   |
| caesar    | 1   |
| I         | 1   |
| was       | 1   |
| killed    | 1   |
| i'        | 1   |
| the       | 1   |
| capitol   | 1   |
| brutus    | 1   |
| killed    | 1   |
| me        | 1   |
| so        | 2   |
| let       | 2   |
| it        | 2   |
| be        | 2   |
| with      | 2   |
| caesar    | 2   |
| the       | 2   |
| noble     | 2   |
| brutus    | 2   |
| hath      | 2   |
| told      | 2   |
| you       | 2   |
| caesar    | 2   |
| was       | 2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2 |
| ambitious | 2   |
|           |   |
|           |   |
|           |   |

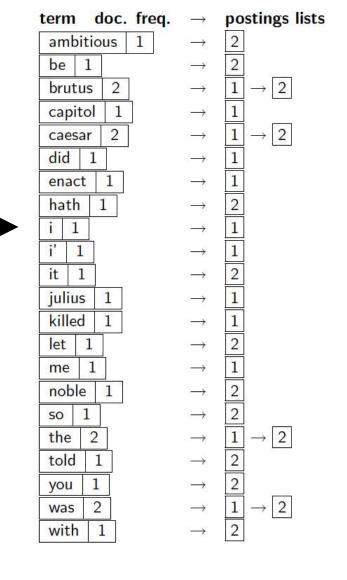
| Term      | docID   |
|-----------|---|
| ambitious | docID 2 2 1 1 2 2 2 2 1 1 1 1 1 1   |
| be        | 2   |
| brutus    | 1   |
| brutus    | 2   |
| capitol   | 1   |
| caesar    | 1   |
| caesar    | 2   |
| caesar    | 2   |
| did       | 1   |
| enact     | 1   |
| hath      | 1   |
| 1         | 1   |
| I a       | 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<br>1<br>1<br>2<br>1<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2<br>2 |
| you       | 2   |
| was       | 1   |
| was       | 2   |
| with      | 2   |
|           |   |

## Indexer steps: Dictionary & Postings

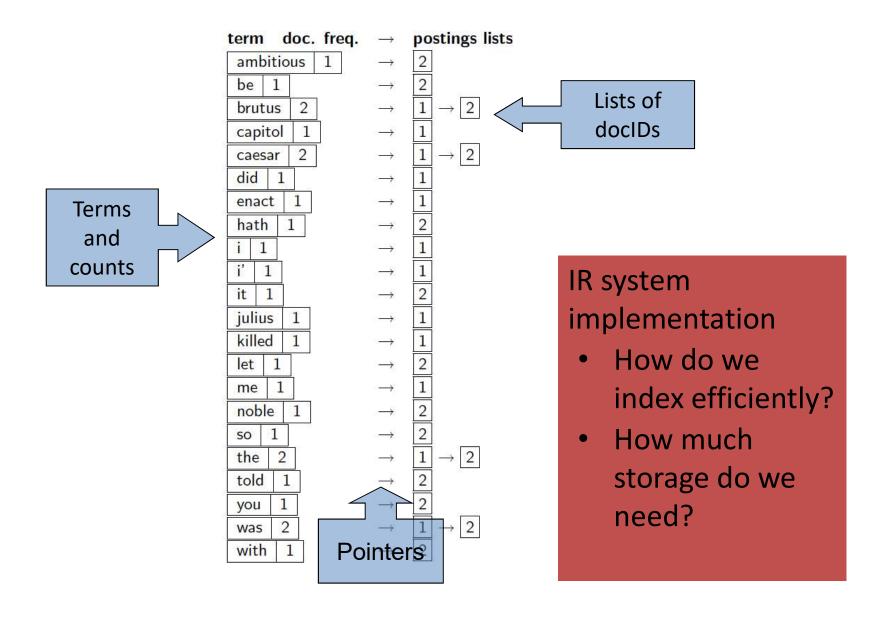
- Multiple term entries in a single document are merged.
- Split into Dictionary and Postings
- Doc. frequency information is added.







# 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?



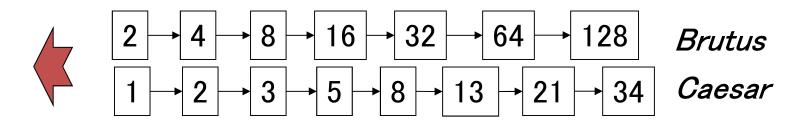
– Later - what kinds of queries can we process?

## Query processing: AND

Consider processing the query:

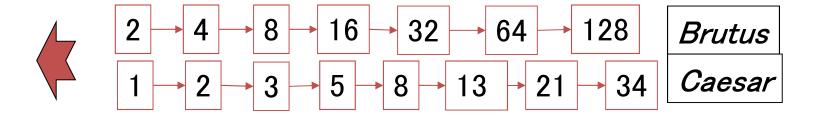
#### **Brutus** AND Caesar

- Locate *Brutus* in the Dictionary;
  - Retrieve its postings.
- Locate *Caesar* in the Dictionary;
  - Retrieve its postings.
- "Merge" the two postings (intersect the document sets):



## 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.

<u>Crucial</u>: postings sorted by docID.

# Intersecting two postings lists (a "merge" algorithm)

```
INTERSECT(p_1, p_2)
       answer \leftarrow \langle \rangle
     while p_1 \neq \text{NIL} and p_2 \neq \text{NIL}
  3
      do if docID(p_1) = docID(p_2)
              then ADD(answer, doclD(p_1))
                      p_1 \leftarrow next(p_1)
  5
                      p_2 \leftarrow next(p_2)
  6
              else if doclD(p_1) < doclD(p_2)
                         then p_1 \leftarrow next(p_1)
  8
                         else p_2 \leftarrow next(p_2)
  9
 10
       return answer
```

# Introduction to Information Retrieval

The Boolean Retrieval Model

& Extended Boolean Models

#### Sec. 1.3

## 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?

#### Sec. 1.3

## Merging

What about an arbitrary Boolean formula?

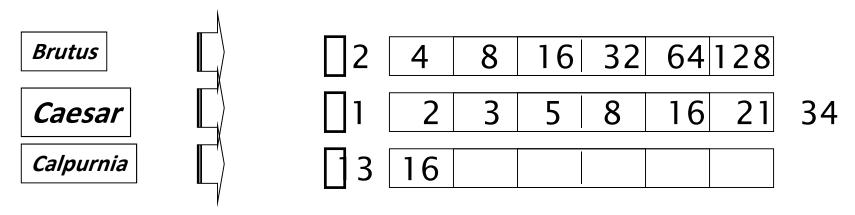
(Brutus OR Caesar) AND NOT

(Antony OR Cleopatra)

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

# Query optimization

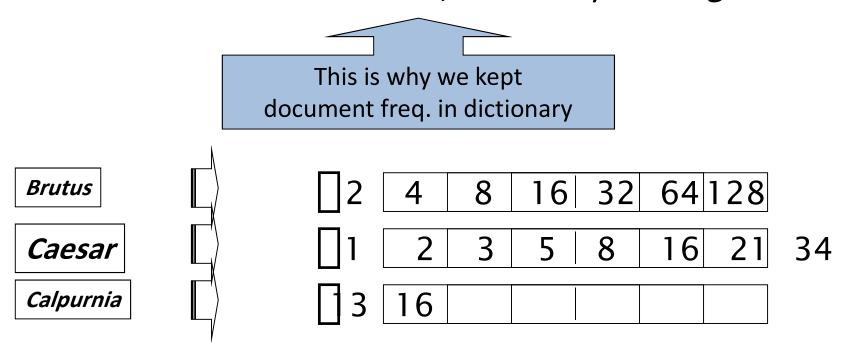
- What is the best order for query processing?
- Consider a query that is an AND of n terms.
- For each of the *n* terms, get its postings, then *AND* them together.



Query: Brutus AND Calpurnia AND Caesar

# Query optimization example

- Process in order of increasing freq:
  - 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 <a href="http://www.rhymezone.com/shakespeare/">http://www.rhymezone.com/shakespeare/</a>
- 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

```
<be: 993427;
1: 7, 18, 33, 72, 86, 231;
2: 3, 149;
4: 17, 191, 291, 430, 434;
5: 363, 367, ...>
Which of docs 1,2,4,5
could contain "to be
or not to be"?
```

- 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 //R

## 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
  - Average web page has <1000 terms</li>
  - SEC filings, books, even some epic poems ... easily 100,000 terms

Consider a term with frequency 0.1%

|               |          | V ( / 1 / 1 / 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 Object Oriented Programming AND Author something like stro\*rup
  - where \* is the wild-card operator