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Text Technologies for Data Science

INFR11145

Indexing

Instructor:
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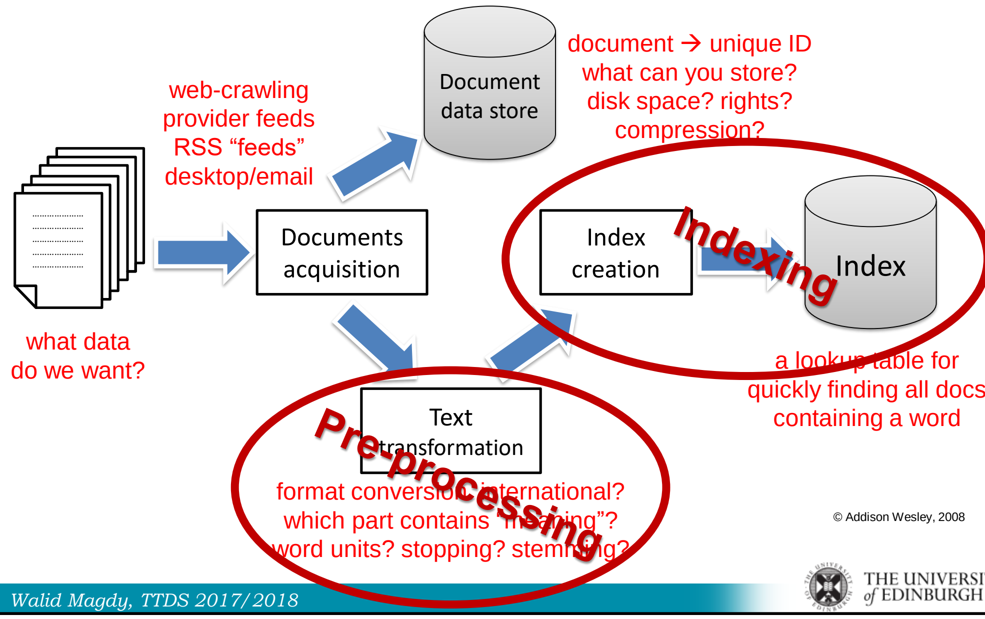
Lecture Objectives

- Learn about and implement
- Boolean search
- Inverted index
- Positional index

Indexing Process



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Pre-processing output

This is an **example** **sentenc**e of how the **pre-process**ing is **appli**ed to **text** in **inform**ation **retriev**al. It **includ**es: **Token**ization, **Stop Word**s **Remov**al, and **Stem**ming

exampl sentenc pre process appli text inform retriev includ token stop word remov stem

- Add processed terms to index
- What is "index"?

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Book Index

Index

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Indexing

- Search engines vs PDF find or grep?
  - Infeasible to scan large collection of text for every "search"
- Book Index
  - For each word, list of "relevant" pages
  - Find topic in sub-linear time
- IR Index:
  - Data structure for fast finding terms
  - Additional optimisations could be applied

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Document Vectors

- Represent documents as vectors
  - Vector → document, cell → term
  - Values: term frequency or binary (0/1)
  - All documents → collection matrix

he	drink	ink	likes	pink	think	wink	
2	1	0	2	0	0	1	← **D1:** He likes to wink, he likes to drink
1	3	0	1	0	0	0	← **D2:** He likes to drink, and drink, and drink
1	1	1	1	0	1	0	← **D3:** The thing he likes to drink is ink
1	1	1	1	1	0	0	← **D4:** The ink he likes to drink is pink
1	1	1	1	1	0	1	← **D5:** He likes to wink, and drink pink ink

Inverted Index

- Represent terms as vectors
  - Vector → term, cell → document
  - Transpose of the collection matrix
  - Vector: inverted list

he	drink	ink	likes	pink	think	wink	
2	1	0	2	0	0	1	← **D1:** He likes to wink, he likes to drink
1	3	0	1	0	0	0	← **D2:** He likes to drink, and drink, and drink
1	1	1	1	0	1	0	← **D3:** The thing he likes to drink is ink
1	1	1	1	1	0	0	← **D4:** The ink he likes to drink is pink
1	1	1	1	1	0	1	← **D5:** He likes to wink, and drink pink ink

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Boolean Search

- Boolean: exist / not-exist
- Multiword search: logical operators (AND, OR, NOT)
- Example
  - Collection: search Shakespeare's Collected Works
  - Boolean query: Brutus AND Caesar and NOT Calpurnia
- Build a **Term-Document Incidence Matrix**
  - Which term appears in which document
  - Rows are terms
  - Columns are documents

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Collection Matrix

Documents

	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

Terms

1 if ***document*** contains ***term***, 0 otherwise

Query: Brutus AND Caesar and NOT Calpurnia
Apply on rows: **110100** AND **110111** AND !(**010000**) = **100100**

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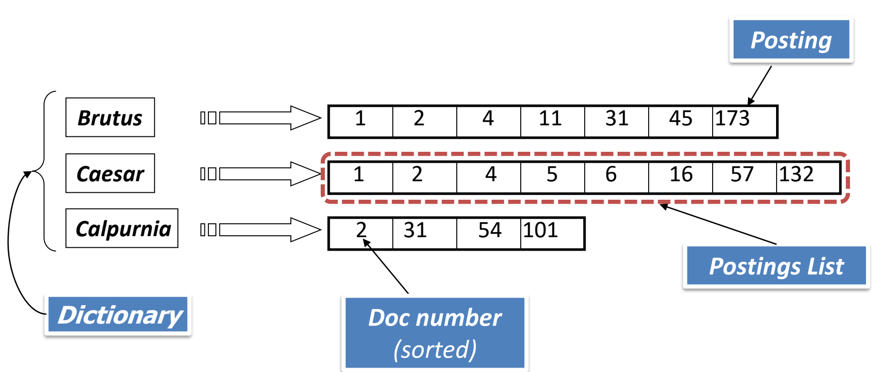
Bigger collections?

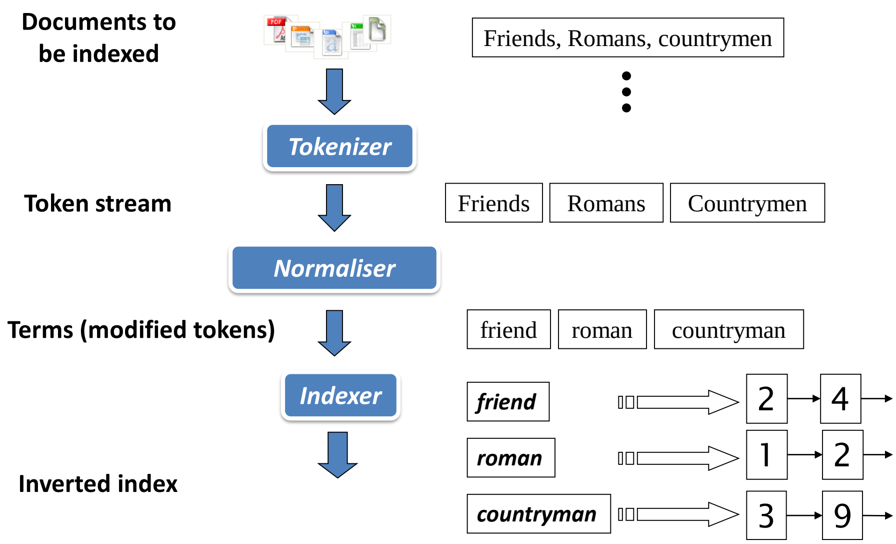
- Consider *N* = 1 million documents, each with about 1000 words.
- *n* = 1M x 1K = 1B words
  ➔ Heap's law → *v* ≈ 500K
- Matrix size = 500K unique terms x 1M documents = 0.5 trillion 0's and 1's entries!
- If all words appear in all documents
  → max{count(1's)} = N * doc. length = 1B
- Actually, from Zip's law → 250k terms appears once!
- Collection matrix is extremely **sparse**.

Inverted Index: Sparse representation

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



Inverted Index Construction



Step 1: Term Sequence

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

Sequence of (term, Doc ID) pairs

Term	docID
I	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
ambitious	2

Step 2: Sorting

- **Sort by:**
  - **1) Term**
  - **then**
  - **2) Doc ID**

Term	docID
I	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
ambitious	2

Sorting →

Term	docID
ambitious	2
be	2
brutus	1
brutus	2
capitol	1
caesar	1
caesar	2
caesar	2
did	1
enact	1
hath	1
I	1
I	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	2
with	2

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Step 3: Posting

1. Multiple term entries in a single document are merged
2. Split into Dictionary and Postings
3. Doc. Frequency (*df*) information is added

Term	docID
ambitious	2
be	2
brutus	1
brutus	2
capitol	1
caesar	1
caesar	2
caesar	2
did	1
enact	1
hath	1
I	1
I	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	2
with	2

→

term	doc. freq.	→	postings lists
ambitious	1	→	2
be	1	→	2
brutus	2	→	1 → 2
capitol	1	→	1
caesar	2	→	1 → 2
did	1	→	1
enact	1	→	1
hath	1	→	2
i	1	→	1
i'	1	→	1
it	1	→	2
julius	1	→	1
killed	1	→	1
let	1	→	2
me	1	→	1
noble	1	→	2
so	1	→	2
the	2	→	1 → 2
told	1	→	2
you	1	→	2
was	2	→	1 → 2
with	1	→	2

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Inverted Index: with frequency

- Boolean: term → DocIDs list
- Frequency: term → touples (DocID,count(term)) lists

he	1:2	2:1	3:1	4:1	5:1
drink	1:1	2:3	3:1	4:1	5:1
ink	3:1	4:1	5:1		
pink	4:1	5:1			
thing	3:1				
wink	1:1	5:1			

appeared in
D2 3 times

Query Processing

- Find documents matching query {ink **AND** wink}
  1. Load inverted lists for each query word
  2. Merge two postings lists → **Linear merge**
- Linear merge → O(*n*)
  n: total number of posts for all query words

ink	3:1	4:1	5:1
wink	1:1	5:1	

Matches

1: f(0,1)

3: f(1,0)

4: f(1,0)

5: f(1,0)

Phrase Search

- Find documents matching query “pink ink”
  1. Find document containing both words
  2. Both words has to be a phrase
- Bi-gram Index:

He likes to wink, and drink pink ink → Convert to bigrams

He_likes likes_to to_wink wink_and and_drink drink_pink pink_ink

- Bi-gram Index, issues:
  - Fast, but index size will explode!
  - What about trigram phrases?
  - What about proximity? “ink is pink”

Proximity Index

- Terms positions is embedded to the inv. Index
  - Called proximity/positional index
  - Enables phrase and proximity search
  - Toubles (DocID, term position)

D1: He likes to wink, he likes to drink

D2: He likes to drink, and drink, and drink

D3: The thing he likes to drink is ink

D4: The ink he likes to drink is pink

D5: He likes to wink, and drink pink ink

he ⇨ 1:2 | 2:1 | 3:1 | 4:1 | 5:1

drink ⇨ 1:1 | 2:3 | 3:1 | 4:1 | 5:1

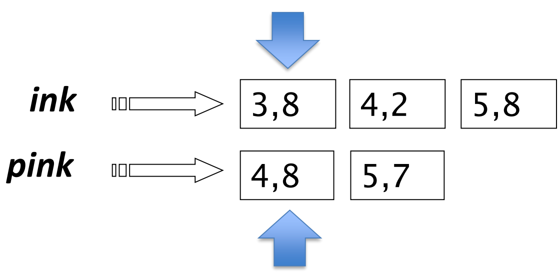
⇩

he ⇨ 1,1 | 1,5 | 2,1 | 3,3 | 4,3 | 5,1

drink ⇨ 1,8 | 2,4 | 2,6 | 2,8 | 3,6 | 4,5 | 5,6

Query Processing: Proximity

- Find documents matching query "pink ink"
  1. Use **Linear merge**
  2. Additional step: check terms positions
- **Proximity search**:
  pos(term1) – pos(term2) < |*w*| → #5(pink,ink)



Matches

3: f(1,0) = **0**

4: f(1,1) = ? =
pos(ink) – pos(pink) == 1?

$\boxed{\text{5: f(1,1) = ? =}}$
pos(ink) – pos(pink) == 1?

Proximity search: data structure

- Possible data structure:
  <term: df;
  DocNo: pos1, pos2, pos3
  DocNo: pos1, pos2, pos3
  ……. >
- Example:
  <***be***: 993427;
  1: 7, 18, 33, 72, 86, 231;
  2: 3, 149;
  4: 17, 191, 291, 430, 434;
  5: 363, 367, …>

Resources

- Text book 1: Intro to IR, Chapter 1 & 2.4
- Text book 2: IR in Practice, Chapter 5