



DD2476 Search Engines and Information Retrieval Systems

Lecture 1: Introduction

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<https://www.kth.se/social/course/DD2476/>

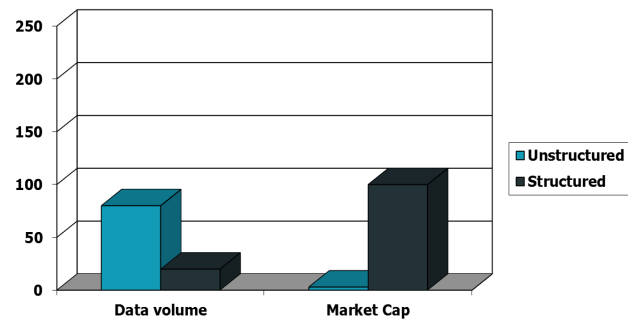
Definition

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).

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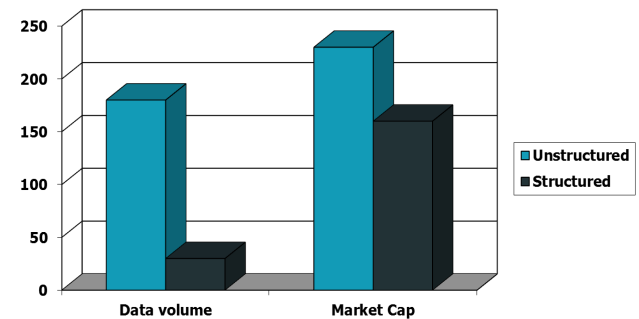
Unstructured (text) vs structured (database) data in the mid-nineties



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Unstructured (text) vs structured (database) data today



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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 in
Lecture 3

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Today

Presentation of lecturers

Course practicalities

- Curriculum
- Examination
- Course homepage:
<https://www.kth.se/social/course/DD2476>

Boolean retrieval (Manning Chapter 1)

- Building an inverted index
- Boolean queries

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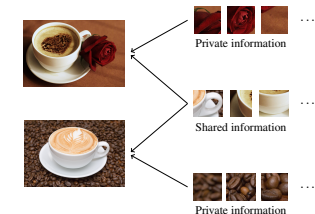


Presentation of Lecturers



Hedvig Kjellström

Associate Professor at CSC
Researcher in Robotics at [CVAP](#), CSC
Lecture 1, 5, 6, 7



Zhang et al, CVPR subm 2016

Figure 1. An example of modeling "a cup of coffee" images. Different images with a cup of coffee all share certain patterns, such as cup handles, cup brims, etc. Moreover, each image also contains patterns that are not immediately related to the "cup of coffee" label, such as the rose or the coffee beans. They can be thought of as private for each image, or instance-specific.

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Johan Boye

Associate Professor at CSC
Researcher in Language
Technology at [TMH](#), CSC
Lecture 2



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Jussi Karlgren

Founder of Gavagai AB, Adjunct
Professor at CSC
Researcher in Language
Technology at [TMH](#), CSC
Lecture 3, 4



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Viggo Kann

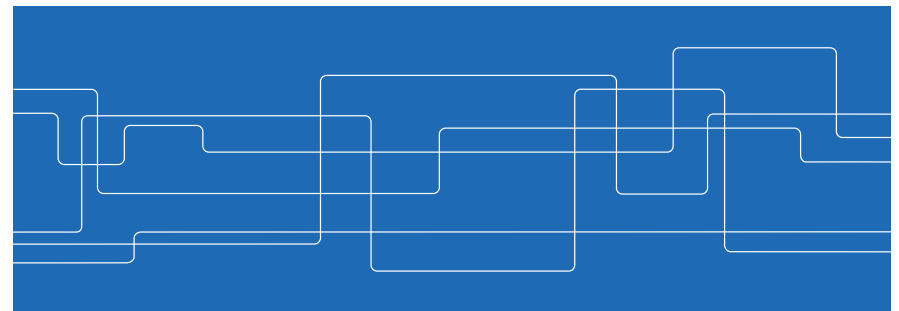
Professor at CSC
Researcher in Theoretical
Computer Science at [TCS](#), CSC
Lecture 8



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Course Practicalities



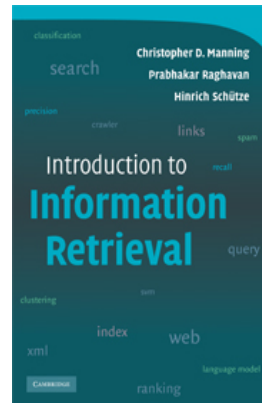


Curriculum

C. D. Manning, P. Raghavan and H. Schütze, *Introduction to Information Retrieval*, Cambridge University Press, 2008

Preliminary version available online in pdf format

- See course homepage: <https://www.kth.se/social/course/DD2476>



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Curriculum

The major part of the book will be covered

Depth according to learning outcomes

- See course homepage: <https://www.kth.se/social/course/DD2476>

Reading on your own necessary

- Lectures cover only highlights, very high pace
- Examination on whole curriculum

Course given for the 7th time

- Changed the evaluation tasks in the assignments

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Curriculum

Field moving forward at high pace

This course a foundation to enable you to learn for yourself

Source: Annual conference ACM SIGIR

Assignments: Give basics (**turn-of-the-century** search engine)

Project: Chance to have a glimpse of the state-of-the-art

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Examination

Three computer assignments (6 ECTS, A-F)

- Individually
- Lab 1 (Lecture 1-3 readings) [February 9](#)
- Lab 2 (Lecture 4-6 readings) [March 8](#)
- Lab 3 (Lecture 7-8 readings) [April 1](#)

Project (3 ECTS, A-F)

- Groups of four-five students
- Presentation (Whole curriculum) [May 20](#)

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Course Homepage

News!

Schedule with readings and examination deadlines

Contact information

Computer assignment and project descriptions

<https://www.kth.se/social/course/DD2476>

Set it to send you email!

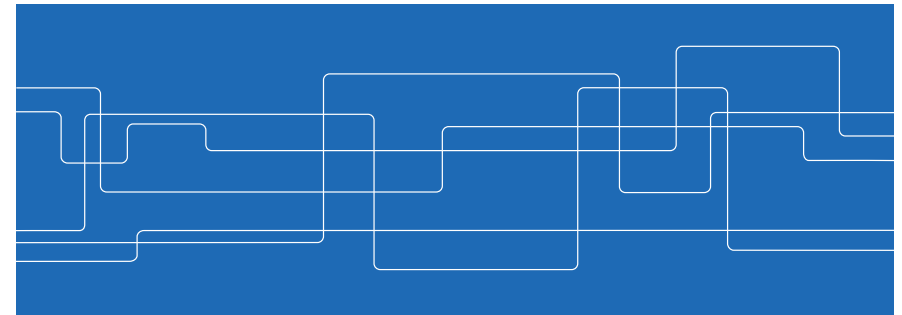


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Boolean Retrieval

(Manning Chapter 1)



A First Information Retrieval Example

Ad hoc retrieval: Find documents in a **collection** of documents (**corpus**), relevant to a certain user need

Boolean retrieval model: Model in which queries are posed as Boolean expressions

Example: Shakespeare

- Find all Shakespeare plays that contain the words

BRUTUS AND CAESAR AND NOT CALPURNIA



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BRUTE Force Approach

One could **grep** all of Shakespeare's plays for BRUTUS and CAESAR, then strip out plays containing CALPURNIA

- Unix command **grep**, linear search

Why is that not the answer?

- Slow (for large corpora)
- Other operations (e.g., find the word ROMANS NEAR COUNTRYMEN) not feasible
- Ranked retrieval (best documents to return)

Instead, organize beforehand

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Term-Document Incidence Matrix

Document = play

	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

Term = word

1 if play contains word, 0 otherwise

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Bitwise Operations

Document = play

	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

Term = word

BRUTUS AND CAESAR AND NOT CALPURNIA
110100 AND 110111 AND NOT 010000
110100 AND 110111 AND 101111
= 100100 (Antony and Cleopatra, Hamlet)

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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.



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Exercise 5 Minutes

Consider 10^6 documents, each with $\sim 10^3$ words.
Avg 6 bytes/word including spaces/punctuation
• 6GB of data.
Say there are $0.5 \cdot 10^6$ *distinct* terms among these.
Normal size collection!

Discuss in pairs:

- What are the problems with using the term-document incidence matrix on a collection this size?
- How can the method be adapted to solve these problems?

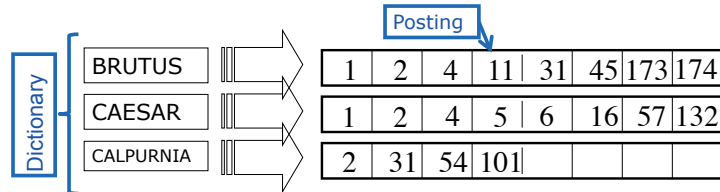
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Inverted Index

For each term t , store a list of all documents that contain t .

- Identify each by a **docID**, a document serial number



Can we use fixed-size arrays for this?

What happens if the term CAESAR is added to document 14?

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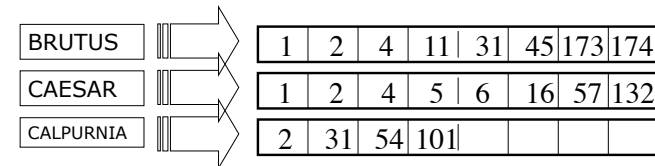


Inverted Index

Need variable-size posting lists

Implementational details

- trade-off storage size/ease of insertion
- Sort lists wrt DocID

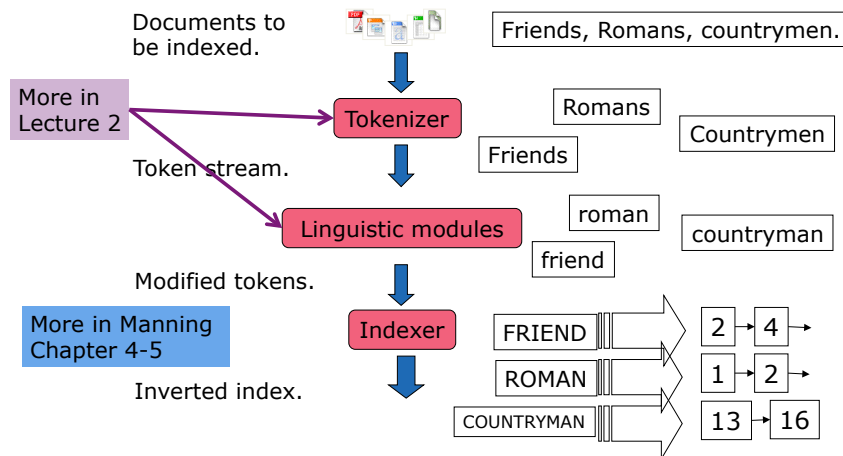


More in Manning
Chapter 4-5

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Building an Inverted Index

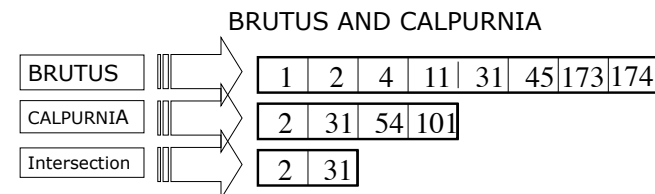


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Query Processing with Inverted Index

Boolean queries are processed as with the incidence matrix



NOT can also be handled with search
Organizing this work (sorting, evaluation order):
[query optimization](#)

More in Manning
Chapter 1

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Beyond Term Search

Allow compounds, e.g., phrases "..."

- "FRIENDS, ROMANS, COUNTRYMEN!"

More in
Lecture 2

Additional operators, e.g., NEAR

- CAESAR NEAR CALPURNIA
- Index has to capture term proximity

Zones in documents

- (author = SHAKESPEARE) AND (text contains WORSER)

More in Manning
Chapter 10

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Beyond Term Search

Not only presence/absence, but also **term frequency**

- 0 vs 1 hit
- 1 vs 2 hits
- 2 vs 3 hits
- Usually, more is better

More in
Lecture 5

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Exercise 5 Minutes

Try the search feature at

www.rhymezone.com/shakespeare

- Who has an open browser? Find someone nearby, or come up to me.

Discuss in groups:

- What could it do better?
- Write down

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IR vs Databases: Structured vs Unstructured Data

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

Typically allows numerical range and exact match (for text) queries, e.g.,

Salary ≥ 60000 AND *Manager* = Smith.

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Unstructured Data

More in
Lectures
9-12

Typically refers to free text but could also be

- Images
- Other media files

More in
Lecture 5

Allows

- Keyword queries
- Free text queries e.g., find all web pages dealing with "drug abuse"
- Classic model for searching text documents

More in
Lecture 7

No data is truly unstructured

- Grammar
- Semistructured search, e.g., XML

More in Manning
Chapter 10

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Organizing Data

More in
Manning
Chapter
16-17

Boolean queries only give inclusion or exclusion of docs.

Clustering: Given a set of docs, group them into clusters based on their contents.

More in
Manning
Chapter
13-14

Classification: Given a set of topics, plus a new doc D , decide which topic(s) D belongs to.

More in
Lecture 5

Ranking: Can we learn how to best order a set of documents, e.g., a set of search results

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The Web and Its Challenges

Unusual and diverse documents

Unusual and diverse users, queries, information needs

Beyond terms, exploit ideas from social networks

- E.g. link analysis

More in
Lecture 6

How do search engines work? And how can we make them better?

More in Lectures
6, 9-12

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Next

Lecture 2 (January 22, 10.15-12.00)

- D3
- Readings: Manning Chapter 2, 3

Computer Assignment 1 (now – February 9)

- Assignment description:

<https://www.kth.se/social/course/DD2476>

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