# INFORMATION RETRIEVAL

Laura Nenzi Inenzi@units.it

Many Slides made by Luca Manzoni

Lecture 1

### WHO AM I

#### AND SOME INFORMATION ABOUT THE COURSE

Laura Nenzi

email: Inenzi@units.it

office: C3, 2nd floor, room 2.55



All the material of the course will be available online on Moodle and Teams

### BOOK



Christopher D. Manning Prabhakar Raghavan Hinrich Schütze

Introduction to Information Retrieval Cambridge University Press. 2008.

Freely available at: <a href="https://nlp.stanford.edu/IR-book/">https://nlp.stanford.edu/IR-book/</a>

### THE EXAM (1)

How

Project + presentation of the project (with questions)

#### When

Exam sessions The projects should be assigned in advance (end of May)

Is the project the same for everybody? No, I admit a maximum number of 3 people for project

How can we select a project? There will be a selection of project to choose from or you can propose your own

### THE EXAM (2)

What kind of project? Code or report (approx. 10 pages)

Can I use language X?

Generally yes, but please ask if it is not in the following list: Python, C, C++, Java, R

> I want to do a project on X We can discuss and prepare a personalised project on X

How is the final mark computed? It is the average of the marks in the two parts

### OUTLINE OF THE COURSE INFORMATION RETRIEVAL PART

- Introduction to information retrieval (IR)
- Data structures for IR
- Models for IR:
  - Boolean
  - Vector space
  - Probabilistic
- Evaluation of IR
- IR on the Web
- Recommender Systems

### LECTURE OUTLINE

Introduction to information retrieval

**Boolean Retrieval** 

Basic Operation on inverted indices



Postings list

Tokenization, Stop Words, Normalization, Stemming & Lemmatization 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)

> Manning, Raghavan, Schütze Introduction to Information Retrieval

(IR) part of computer science which studies the retrieval of information (not data) from a collection of written documents. The retrieved documents aim at satisfying a user information need usually expressed in natural language. Baeza-Yates, Ribeiro-Neto **Modern Information Retrieval** 

### **EXAMPLES OF INFORMATION RETRIEVAL**

- Web search
- Searching emails in your email client
- Searching documents using
  Spotlight/Windows Desktop Search/Tracker/Nepomuk/Baloo
- Search inside a knowledge base

### THREE PROMINENT SCALES

- web search: over billions of documents stored on millions of computer
- personal information retrieval: searching emails in your email client, spotlight/Windows Desktop Search/Tracker/Nepomuk/ Baloo
- enterprise, institutional, and domain-specific search: collections such as a corporation's internal documents, a database of patents, or research articles on biochemistry

#### A BRIEF HISTORY OF IR BEFORE COMPUTERS

The origin of IR is in the cataloguing of books in libraries

Each books has one or more topics associated to it

Multiple machines were invented to help librarians in retrieving books



-e 19.3. GE MN

Inventor: Enance Greater

MACHINE FOR RETRIEVAL BASED ON MICROFILM AND A PHOTOCELL. GOLDBERG, 1931

Sanderson, Mark, and W. Bruce Croft.

"The history of information retrieval research."

Proceedings of the IEEE 100.Special Centennial Issue (2012): 1444-1451.

### A BRIEF HISTORY OF IR EARLY USE OF COMPUTERS

The use of computer allowed to speed up the traditional search methods, like searching across 10<sup>6</sup> records in 15 hours using a UNIVAC computer

But new techniques were also being developed:

First move to indexing based on words instead of assigning topics to document (*Uniterm system, 1952*)



How to rank the results: position based on the frequency of the words in the documents (term frequency weighting, 1958)

#### A BRIEF HISTORY OF IR PRE-INTERNET ('60-EARLY '90)

**Clustering** of documents with similar contents

Representing documents as vectors: the vector space mode

Relevance feedback: the user feedback can be used to improve the query results

First private companies focused on information retrieval

tf-idf: less common words tend to refer to more specific concepts, which were more important in retrieval

Text REtrieval Conference (TREC)

#### A BRIEF HISTORY OF IR WEB SEARCH

1993: ~100 websites 2019: ~2 billions websites

Instead of using an authoritative source now the web must be scraped to get information

The links between the pages are a source of additional information



Altavista: first fully-searchable index of the web (1995)



Google: currently the most used web search engine (1997)

### SOME TERMINOLOGY DOCUMENTS AND COLLECTIONS

- Document: individual unit on which we build an IR system
  - Books, chapters, webpages, scientific papers, etc.
  - Images, videos, Music, etc.
- Collection (also called Corpus): the group of documents on which we perform the search
  - All Shakespeare plays, the emails in your mailbox, all the products on an online shop, the web
  - Some collections are static, while others are dynamic

### 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

### PROPERTIES OF UNSTRUCTURED DOCUMENTS

#### DOCUMENT

- Significant text contents
- Some structure might be present (e.g., title, author)
- The semantics of the document
  is not well-defined
- Example query: "find all stories about holidays in Ireland"

#### DATABASE RECORD

- Predefined structure
- Usually the semantics of the fields is well defined
- Finding matches is (usually) done by comparing specific fields
- Example Query: "select all products that cost at most 50€"

### SOME TERMINOLOGY (2) FROM INFORMATION NEED TO QUERIES

- Information need: a topic about which the user what to know more
- Query: the way the user formulate his/her information need to the IR system
  - For the same information need the users might formulate different queries
  - E.g., what query would you use to know which is the current record holder in Tetris?

### GOAL OF AN IR SYSTEM

An IR System must interpret the information needs of the user and estimate the relevance of the documents with respect to it.



#### IS AN IR SYSTEM JUST "GREP"? HINT: NO

> grep term\_to\_search corpus



- Searching across all the text does not scale to large collections
- We might want a more flexible query language
- We might want ranked retrieval (i.e., more relevant documents first)

### INDEXING HOW TO AVOID GREPPING

- It is unfeasible to scan the entire corpus for each query...
- ...therefore, the corpus is scanned once\* and an index is built
- An index will have for each **term** (e.g., a word) all documents containing that term
- In this way we can avoid a linear scan of all documents

### STRUCTURE OF AN IR SYSTEM



### MAIN COMPONENTS

A FORMAL REPRESENTATION OF THE QUERIES

**Exact matching** The document is either relevant or not relevant

A WAY OF MATCHING THE QUERY AND DOCUMENTS REPRESENTATIONS AND A WAY TO MEASURE RELEVANCE

Partial matching The document that are "sufficiently similar" to the query are recovered

A FORMAL REPRESENTATION OF THE DOCUMENTS

### **DIFFERENT KINDS OF REPRESENTATIONS** QUERIES AND DOCUMENTS



### THE TWO ASPECTS OF IR

We need to manage technical and a semantic aspects

#### **Technical aspects**

- How to represent information in a computer?
- How to retrieve the information fast enough?

EFFICIENCY

Semantic aspects

- How can we store information in a way that preserves its meaning?
- How can we retrieve the relevant documents?

**EFFECTIVENESS** 

# **MEASURING EFFECTIVENESS**



$$precision = \frac{relevant \cap retrieved}{retrieved}$$

Which fraction of the retrieved documents is relevant

$$recall = \frac{relevant \cap retrieved}{relevant}$$

Which fraction of the relevant documents has been retrieved



### WHAT IS BOOLEAN RETRIEVAL?

- Only exact matching: either a document is relevant or not
- The query is expressed like a Boolean formula:
  - E.g., (dog OR cat) AND box
- We can ask for the inclusion (or exclusion) of certain terms
- Going forward, we can decide to extend this model to allow more powerful queries
- For now we ignore the issue of ranking the results according to relevance.

### A VISUAL EXAMPLE BOOLEAN QUERIES AND SETS



### LINKING BOOLEAN QUERIES AND SETS

 $d_1 = \{t_1, t_3, t_4\}$  $d_2 = \{t_1, t_2, t_5\}$  $d_2 = \{t_2, t_3, t_5\}$ 

Documents as sets of terms  $t_1 = \{d_1, d_2\}$   $t_2 = \{d_2, d_3\}$   $t_3 = \{d_1, d_3\}$   $t_4 = \{d_1\}$  $t_5 = \{d_2, d_3\}$ 

 $t_2 = \{d_2, d_3\}$ Terms as sets $t_3 = \{d_1, d_3\}$ of documents

Queries	Answers
$q_1 = t_2$	$\{d_2, d_3\}$
$q_2 = t_2 \wedge t_3  \longrightarrow $	$\{d_2, d_3\} \cap \{d_1, d_3\} = \{d_3\}$
$q_3 = t_2 \lor t_3  \longrightarrow $	$\{d_2, d_3\} \cup \{d_1, d_3\} = \{d_1, d_2, d_3\}$

### HOW TO ASSOCIATE DOCUMENTS AND TERMS

Let us consider as a corpus a set of ~400 articles from "Time"<sup>1</sup> in the '60s:

- **O.** THE ALLIES AFTER NASSAU IN DECEMBER 1960, THE U.S . FIRST [...]
- 1. RUSSIA WHO'S IN CHARGE HERE ? IT WAS IN 1954 THAT NIKITA [...]
- **2.** BERLIN ONE LAST RUN HANS WEIDNER HAD BEEN HOPING FOR MONTHS TO [...]
- **3.** THE ROAD TO JAIL IS PAVED WITH NONOBJECTIVE ART SINCE THE [...]

We can build an incidence matrix of documents and terms

<sup>1</sup> Available at <u>http://ir.dcs.gla.ac.uk/resources/test\_collections/time/</u>

# THE INCIDENCE MATRIX

	Article 0	Article 1	Article 2	Article 3	Article 4	Article 5
Α	1	1	1	1	1	1
AACHEN	0	0	0	0	0	0
ABABA	0	0	0	0	0	0
ABABAS	0	0	0	0	0	0
ABACK	0	0	0	0	0	0
ABADAN	0	0	0	0	0	0
ABANDON	0	0	0	0	0	0
ABANDONED	0	0	0	0	0	0
ABANDONING	0	0	0	0	0	0
ABANDONMENT	0	0	0	0	0	0

...and another 22484 rows

### QUERIES

 Query 1 = FRANCE AND BRITAIN

 FRANCE
 1
 0
 0
 0
 0
 0
 0
 1

 BRITAIN
 1
 0
 0
 0
 0
 0
 1
 0
 1

 Document matching Query 1
 Image: Colored and the second and the sec

Query 2 = FRANCE OR BRITAIN

FRANCE	1	0	0	0	0	0	0	0	1	
BRITAIN	1	 0	0	 0	0	0	 1	0	1	

0 0 0 0 1 0

Documents matching Query 2

Query 1

### **INCIDENCE MATRIX** STORAGE REQUIREMENTS

The size needed to store an incidence matrix is #terms × #documents

Hence the size requirements make this data structure impractical

#### A simple example

The Oxford English Dictionary has over 171000 words in common use

In 2019 the English Wikipedia contains over  $5.9 \times 10^6$  articles

The resulting incidence matrix would have more than  $10^{12}$  entries

# THE INCIDENCE MATRIX

	Article 0	Article 1	Article 2	Article 3	Article 4	Article 5
Α	1	1	1	1	1	1
AACHEN	0	0	0	0	0	0
ABABA	0	0	0	0	0	0
ABABAS	0	0	0	0	0	0
ABACK	0	0	0	0	0	0
ABADAN	0	0	0	0	0	0
ABANDON	0	0	0	0	0	0
ABANDONED	0	0	0	0	0	0
ABANDONING	0	0	0	0	0	0
ABANDONMENT	0	0	0	0	0	0

...and another 22484 rows

# INVERTED INDEX: UNION AND INTERSECTION

#### THE INVERTED INDEX A MORE COMPACT DATA STRUCTURE

- We want to avoid storing rows that will be mostly empty
- We store, for each term, the list of documents containing it
- This is similar to the difference between adjacency *matrices* and adjacency *lists* for graphs.
- We keep the list ordered to improve performances for union and intersection
- Why inverted index? Because it is not from documents to terms but from terms to documents. Apart from that, it is a "normal" index

# THE INVERTED INDEX

#### A VISUAL REPRESENTATION

Α 0 2 AACHEN 250 List of terms 167 226 229 ABABA ABABAS 229 326 353 23 ABACK 119 ABADAN 22 271 284 ABANDON ABANDONED 187 192 70 ABANDONING 261 ABANDONMENT 359 Ordered list of documents containing the term

### SOME TERMINOLOGY (3) THIS TIME FOR INVERTED INDICES

- **DocID** (Document Identifier): a unique number associated to a document. E.g., consecutive integers.
- Dictionary: the collection of all terms that we have in the inverted index
- Posting list: list of DocIDs associated to a term
- **Posting**: element of the list (different from a simple DocID because it is associated to a term)

### BUILDING AN INVERTED INDEX SORTING AND GROUPING

- For each document we extract the sequence of terms
- We tag each term with the corresponding DocID
- We sort the list of terms extracted from all the documents
- We group together equal terms and we "merge" the posting lists of the two terms

### EXAMPLE OF SORTING AND GROUPING

#### **Document 1**: the cat is inside the box

**Document 2**: the dog is barking at the box

the	1	at	2		at	2
cat	1	barking	2		barking	2
is	1	box	1		box	1,2
inside	1	box	2		cat	1
the	1	cat	1		dog	2
box	1	dog	2		inside	1
the	2	inside	1	1	is	1,2
dog	2	is	1	11	the	1,2
is	2	is	2			
barking	2	the	1			
at	2	the	1			
the	2	the	2	//		
box	2	the	2	/		

# WORST-CASE BOUNDS AND THE REAL WORLD

- But... the space occupied by an inverted index is not necessarily lower than the one of the incidence matrix.
- In the worst case (each document contains all the terms) they both occupy O (#terms × #documents) space.
- In practice most document contains only a small subset of the terms.
- The same reasoning applies to the time complexity of the operations (intersection and union) performed on the set of documents.

# HOW TO IMPLEMENTATION AND OTHER IMPROVEMENTS

- We will spend some time in discussing how to implement and improve the inverted index
- Basic functionality: answer queries of the form
  - term1 AND term2
  - term1 OR term2

# ANSWERING A SIMPLE QUERY

#### A SINGLE WORD QUERY



QUERY

We find the term in the list of terms

We return the associated list of terms

ANSWERING AN "AND" QUERY NOW WITH TWO WORDS



















#### NOW WITH TWO WORDS



**Complexity**: linear in the lengths of the lists



Size of the answer ≤ minimum of the lengths of the lists

ANSWERING A "OR" QUERY WITHOUT DUPLICATES!





















#### WITHOUT DUPLICATES!



Size of the answer ≤ sum of the lengths of the lists

#### IS THAT ALL? HINT: NO



Some terms are not useful: "A" is in all the documents!



We would like to better determine the set of terms in the dictionary and to provide retrieval that is tolerant to spelling mistakes and inconsistent choice of words.

#### BUILDING AN INVERTED INDEX MAJOR STEPS

- 1. Collect the documents to be indexed
- 2. Tokenize the text, turning each document into a list of tokens
- 3. Do linguistic preprocessing, producing a list of normalized tokens, which are the indexing terms
- 4. Index the documents that each term occurs in by creating an inverted index, consisting of a dictionary and postings.

### IS THAT ALL? HINT: NO

- Often useful to search for compounds or phrase: biwords, term1 NEAR term2, "term1 term2", term1\* (wildcards), etc.
- Term frequency information (the number of times a term occurs in a document) in postings lists
- An effective method to order (or "rank") the returned results
- How to compress the index, how to update it, etc.