

# VK Multimedia Information Systems

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Dienstags, 16.00 Uhr s.t., E.1.42



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# Information Retrieval Basics: Agenda



- Information Retrieval History
- Information Retrieval & Data Retrieval
- Searching & Browsing
- Information Retrieval Models
- Exercise 01



# Information Retrieval History



Currently there are no museums for IR

IR is the process of **searching** through a **document collection** based on a particular **information need**.



### **IR Key Concepts**



- Searching
  - o Indexing, Ranking
- Document Collection
  - Textual, Visual, Auditive

- Particular Needs
  - Query, User based







## A History of Libraries



Libraries are perfect examples for document collections.

- Wall paintings in caves
  - e.g. Altamira, ~ 18,500 years old
- Writing in clay, stone, bones
  - e.g. Mesopotamian cuneiforms, ~ 4.000 BC
  - e.g. Chinese tortoise-shell carvings, ~ 6.000 BC
  - e.g. Hieroglyphic inscriptions,
     Narmer Palette ~ 3.200 BC





# A History of Libraries (ctd.)



#### Papyrus

- Specific plant (subtropical)
- Organized in rolls, e.g. in Alexandria

#### Parchment

- Independence from papyrus
- Sewed together in books

#### Paper

- Invented in China (bones and bamboo too heavy, silk too expensive)
- Invention spread -> in 1120 first paper mill in Europe



# A History of Libraries (ctd.)



- Gutenberg's printing press (1454)
  - Inexpensive reproduction
  - o e.g. "Gutenberg Bible"
- Organization & Storage
  - Dewey Decimal System (DDC, 1872)
  - Card Catalog (early 1900s)
  - Microfilm (1930s)
  - MARC (Machine Readable Cataloging, 1960s)
  - Digital computers (1940s+)



### **Library & Archives today**



- Partially converted to electronic catalogues
  - From a certain time point on (1992 ...)
  - Often based on proprietary systems
  - Digitization happens slow
  - No full text search available
  - Problems with preservation
    - Storage devices & formats



### **History of Searching**



- Browsing
  - Like "Finding information yourself"
- Catalogs
  - Organized in taxonomies, keywords, etc.
- Content Based Searching
  - o SELECT \* FROM books WHERE title= \%Search%'
- Information Retrieval
  - Ranking, models, weighting
  - Link analysis, LSA, ...



## **History of IR**



- Starts with development of computers.
- Term "Information Retrieval" coined by Mooers in 1952
- Two main periods (Spark Jones u. Willett)
  - 1955 1975: Academic research
    - Models and Basics
    - Main Topics: Search & Indexing
  - 1975 ... : Commercial applications
    - Improvement of basic methods



## A Challenge: The World Wide Web



- First actual implementation of Hypertext
  - Interconnected documents
  - Linked and referenced
- World Wide Web (1989, T. Berners-Lee)
  - Unidirectional links (target is not aware)
  - Links are not typed
  - Simple document format & communication protocol (HTML & HTTP)
  - Distributed and not controlled



### Some IR History Milestones



- Book "Automatic Information Organization and Retrieval", Gerard Salton (1968)
  - Vector Space Model
- Paper "A statistical interpretation of term specificity and its application in retrieval", Karen Spärck Jones (1972)
  - IDF weighting
  - http://www.soi.city.ac.uk/~ser/idf.html
- Book "Information Retrieval" of C.J. Rijsbergen (1975)
  - Probabilistic Model
  - http://www.dcs.gla.ac.uk/Keith/Preface.html



### Some IR History Milestones



- Paper "Indexing by Latent Semantic Analysis", S. Deerwester, Susan Dumais, G. W. Furnas, T. K. Landauer, R. Harshman (1990).
  - Latent Semantic Indexing
- Paper "Some simple effective approximations to the 2-Poisson model for probabilistic weighted retrieval" Robertsen & Walker (1994)
  - BM25 weighting scheme
- Paper "The Anatomy of a Large-Scale Hypertextual Web Search Engine", Sergey Brin & Larry Page (1998)
  - World Wide Web Retrieval



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### Organizational: References



- in the Library
  - Modern Information Retrieval, Ricardo Baeza-Yates & Berthier Ribeiro-Neto, Addison Wesley
  - Google's Pagerank and Beyond: The Science of Search Engine Rankings, Amy N. Langville & Carl D. Meyer, University Presses of CA
  - Distributed Multimedia Database Technologies supported by MPEG-7 and MPEG-21, Harald Kosch, CRC Press
  - Readings in Information Retrieval, Karen
     Sparck Jones, Peter Willett, Morgan Kaufmann



### Organizational: References



#### WWW

- Skriptum Information Retrieval, Norbert Fuhr, Lecture Notes on Information Retrieval - Univ. Dortmund, 1996. Updated in 2002
- Information Retrieval 2nd Edt., C.J. Rijsbergen, Butterworth, London 1979

#### Through me:

 Lectures on Information Retrieval: Third European Summer-School, Essir 2000 Varenna, Italy, Revised Lectures, Maristella Agosti, Fabio Crestani & Gabriela Pasi (eds.), Lecture Notes in Computer Science, Springer 2000



## Information Retrieval & Data Retrieval



#### **Information Retrieval**

- Information Level
- Search Engine
- Teoma / Google

#### **Data Retrieval**

- Data Level
- Data Base
- Oracle / MySQL

## Information Retrieval & Data Retrieval



Information Retrieval	Data Retrieval
Content Based Search	Search for Patterns and String
Query ambigous	Query formal & unambigous
Results ranked by relevance	Results not ranked
Error tolerant	Not error tolerant
Multiple iterations	Clearly defined result set
Examples	Examples
Search for synonyms	Search for patterns
Bag of Words	SQL Statement

Retrieval is nearly always a combination of both.

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# Information Retrieval Basics: Searching



A **user** has an **information need**, which needs to be **satisfied**.

- Two different approaches:
  - Browsing
  - Searching



## Searching & Browsing

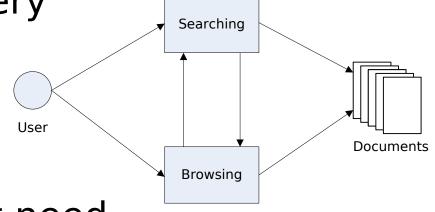


#### Searching

- Explicit information need
- Definition through "query"
- Result lists
- e.g. Google

#### **Browsing**

- Not necessarily explicit need
- Navigation through repositories



### **Browsing**



- Flat Browsing
  - User navigates through set of documents
  - No implied ordering, explicit ordering possible
  - Examples: One single directory, one single file
- Structure Guided Browsing
  - An explicit structure is available for navigation
  - Mostly hierarchical (file directories)
  - Can be generic digraph (WWW)
  - o Examples: File systems, World Wide Web



## Searching



- Query defines "Information Need"
- Ad Hoc Searching
  - Search when you need it
  - Query is created to fit the need
- Information Filtering
  - Make sets of documents smaller
  - Query is filter criterion
- Information Push
  - Same as filtering, delivery is different



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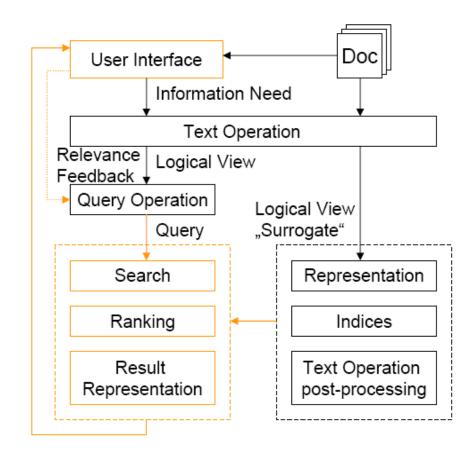


## Information Retrieval System Architecture



#### **Aspects**

- Query & languages
- IR models
- Documents
- Internal representation
- Pre- and post-processing
- Relevance feedback
- HCI



## Information Retrieval Models



- Boolean Model
  - Set theory & Boolean algebra
- Vector Model
  - Non binary weights on dimensions
  - Partial match
- Probabilistic Model
  - Modeling IR in a probabilistic framework

## Formal Definition of Models



## An information retrieval model is a quadruple $[D, Q, F, R(q_i, d_i)]$

- D is a set of logical views (or representations) for the documents in the collection.
- Q is a set of logical views (or representations) for the user needs or queries.
- F is a **framework** for modeling document representations, queries and their relationship.
- $R(q_i, d_j)$  is a **ranking function** which associates a real number with a query  $q_i$  of Q and a document  $d_i$  of D.



#### **Definitions**

#### in Context of Text Retrieval



- index term word of a document expressing (part of) document semantics
- weight  $w_{i,j}$  quantifies the importance of index term  $t_i$  for document  $d_j$
- index term vector for document d<sub>j</sub> (having t different terms in all documents):

$$\overrightarrow{d_j} = (w_{1,j}, w_{2,j}, ..., w_{t,j})$$

### **Boolean Model**

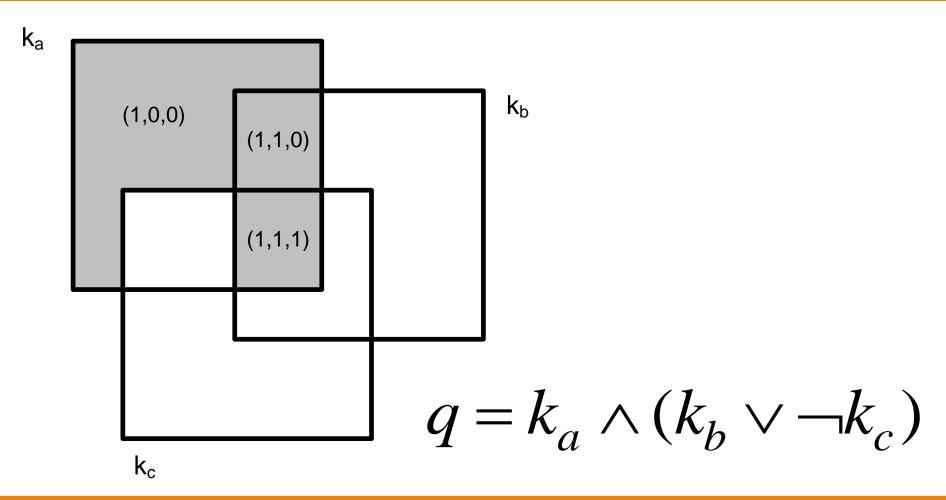


- Based on set theory and Boolean algebra
  - Set of index terms
  - Query is Boolean expression
- Intuitive concept:
  - Wide usage in bibliographic system
  - Easy implementation and simple formalisms
- Drawbacks:
  - Binary decision components (true/false)
  - No relevance scale (relevant or not)



### **Boolean Model: Example**





### **Boolean Model: DNF**



$$q = k_a \land (k_b \lor \neg k_c) \dots \overrightarrow{q}_{dnf} = (1,1,1) \lor (1,1,0) \lor (1,0,0)$$

- Express queries in disjunctive normal form (disjunction of conjunctive components)
- Each of the components is a binary weighted vector associated with  $(k_a, k_b, k_c)$
- Weights  $w_{i,j} \in \{0,1\}$

## Boolean Model: Ranking function



$$sim(d_{j}, q) = \begin{cases} 1 & \text{if } \exists \vec{q}_{cc} \middle| (\vec{q}_{cc} \in \vec{q}_{dnf}) \land (\forall_{k_{i}}, g_{i}(\vec{d}_{j}) = g_{i}(\vec{q}_{cc})) \\ 0 & \text{otherwise} \end{cases}$$

 similarity is one if one of the conjunctive components in the query is exactly the same as the document term vector.



### **Boolean Model**



- Advantages
  - Clean formalisms
  - Simplicity
- Disadvantages
  - Might lead to too few / many results
  - No notion of partial match
  - Sequential ordering of terms not taken into account.



### **Vector Model**



- Integrates the notion of partial match
- Non-binary weights (terms & queries)
- Degree of similarity computed

$$\vec{d}_j = (w_{1,j}, w_{2,j}, ..., w_{t,j})$$

$$\vec{q} = (w_{1,q}, w_{2,q}, ..., w_{t,q})$$



## Vector model: Similarity



$$sim(d_{j}, q) = \frac{\vec{d}_{j} \cdot \vec{q}}{\left| \vec{d}_{j} \right| \times \left| \vec{q} \right|} = \frac{\sum_{i=1}^{t} w_{i, j} \cdot w_{i, q}}{\sqrt{\sum_{i=1}^{t} w_{i, j}^{2} \cdot \sqrt{\sum_{i=1}^{t} w_{i, q}^{2}}}}$$



## Vector Model: Example



$$\vec{0}$$
 = (0.3, 0.4, 0, 0.1, 1)  
 $\vec{0}$  = (1, 0, 0, 0.5, 0)

$$Sim(\vec{0},\vec{9}) = \frac{1.0.3 + 0.1.0.5}{\sqrt{0.3^2 + 0.1^2 + 0.1^2 + 0.1^2 + 1}} \approx \frac{0.35}{2.24} \approx 0.17$$

## **Another Example:**



#### Document & Query:

- D = "The quick brown fox jumps over the lazy dog"
- Q = "brown lazy fox"

$$sim(d_{j},q) = \frac{\vec{d}_{j} \bullet \vec{q}}{\left| \vec{d}_{j} \right| \times \left| \vec{q} \right|} = \frac{\sum_{i=1}^{t} w_{i,j} \cdot w_{i,q}}{\sqrt{\sum_{i=1}^{t} w_{i,j}^{2} \cdot \sqrt{\sum_{i=1}^{t} w_{i,q}^{2}}}}$$

#### Results:

- $\circ (1,1,1,1,1,1,1,1,1)^{t} * (1,1,1,0,0,0,0,0,0)^{t} = 3$
- $\circ$  sqrt(9) \* sqrt(3) = 5,196
- $\circ$  Similarity = 3 / 5,196 = 0,577

## Term weighting: TF\*IDF



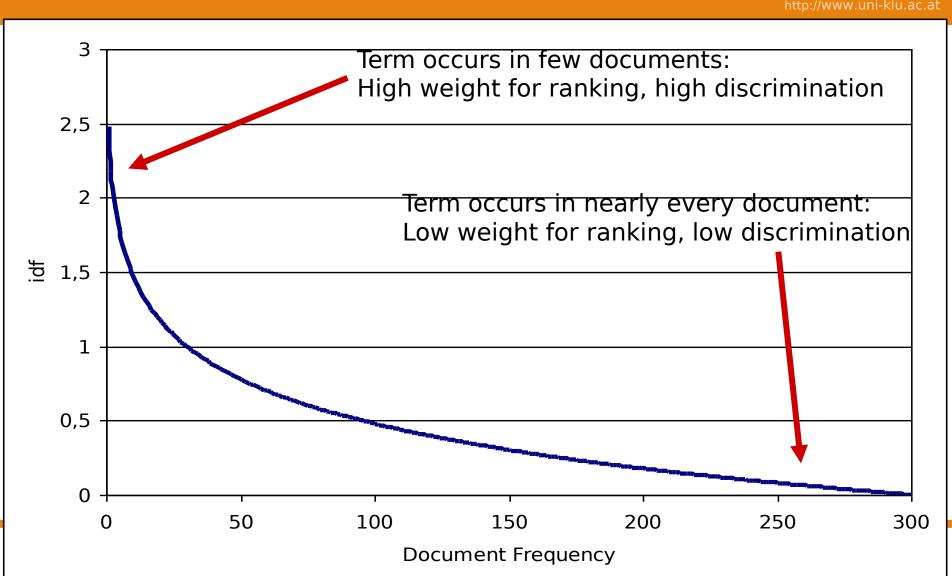
## Term weighting increases retrieval performance ...

- Term frequency
  - Most intuitive approach
  - O How often does a term occur in a document?
- Inverse Document Frequency
  - How much information content has a term for a document collection?
  - Compare to Information Theory of Shannon



## Example: IDF 300 documents corpus





### Definitions: Normalized Term Frequency



$$f_{i,j} = \frac{freq_{i,j}}{\max_{l}(freq_{l,j})} \dots \text{normalized term frequency}$$

 $freq_{i,j}$  ... raw term frequency of term i in document j

- Maximum is computed over all terms in a document
- Terms which are not present in a document have a raw frequency of 0



### Definitions: Inverse Document Frequency



$$idf_i = \log \frac{N}{n_i}$$
 ... inverse document frequency for term  $i$ 

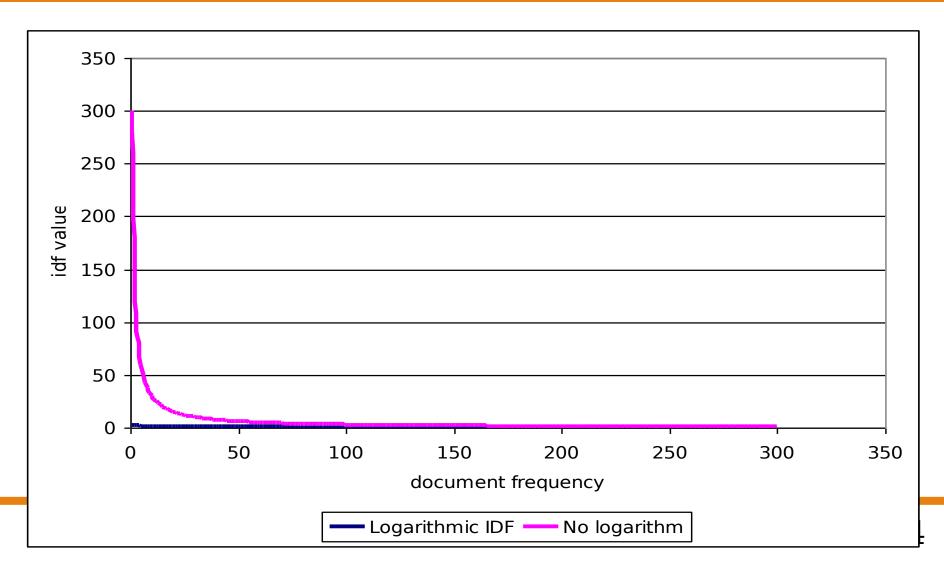
N ... number of documents in the corpus

 $n_i$  ... number of document in the corpus which contain term i

- Note that  $idf_i$  is independent from the document.
- Note that the whole corpus has to be taken into account.

## Why log(...) in IDF?





#### TF\*IDF



- TF\*IDF is a very prominent weighting scheme
  - Works fine, much better than TF or Boolean
  - Quite easy to implement

$$w_{i,j} = f_{i,j} \cdot \log \frac{N}{n_i}$$

## Weighting of query terms



$$w_{i,q} = (0.5 + \frac{0.5 \cdot f_{i,q}}{\max_{l}(f_{l,q})}) \cdot \log \frac{N}{n_i}$$

- Also using IDF of the corpus
- But TF is normalized differently
  - $\circ$  TF > 0.5
- Note: the query is not part of the corpus!

#### **Vector Model**



- Advantages
  - Weighting schemes improve retrieval performance
  - Partial matching allows retrieving documents that approximate query conditions
  - Cosine coefficient allows ranked list output
- Disadvantages
  - Term are assumed to be mutually independent



## Simple example (i)



#### Scenario

- Given a document corpus on birds: nearly each document (say 99%) contains the word bird
- someone is searching for a document about sparrow nest construction with a query "sparrow bird nest construction"
- Exactly the document which would satisfy the user needs does not have the word "bird" in it.



## Simple example (ii)



- TF\*IDF weighting
  - knows upon the low discrimative power of the term bird
  - The weight of this term is near to zero
  - This term has virtually no influence on the result list.



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#### **Exercise**



- Given a document collection ...
- Find the results to a query ...
  - Employing the Boolean model
  - Employing the vector model (with TF\*IDF)
- Some hints:
  - o Excel:
    - Sheet on homepage
    - Use functions "Summenprodukt" & "Quadratesumme"



#### **Exercise**



- Document collection (6 documents)
  - o spatz, amsel, vogel, drossel, fink, falke, flug
  - spatz, vogel, flug, nest, amsel, amsel
  - o kuckuck, nest, nest, ei, ei, ei, flug, amsel, amsel, vogel
  - o amsel, elster, elster, drossel, vogel, ei
  - falke, katze, nest, nest, flug, vogel
  - o spatz, spatz, konstruktion, nest, ei
- Queries:
  - spatz, vogel, nest, konstruktion
  - o amsel, ei, nest



## **Exercise**



	d1	d2	d3	d4	d6	d6	idf
amsel	1	3	2	1			
drossel	1			1			
ei			3	1		1	
elster				2			
falke	1				1		
fink	1						
flug	1	1	1		1		
katze					1		
konstruktion						1	
kuckuck			1				
nest		1	2		2	1	
spatz	1	1				2	
vogel	1	1	1	1	1		

# Participation in a Research Project ?!?



We investigate user behavior in multimedia search in an exploratory analysis ...

- Please take a sheet
- Take some time and go through the tasks
- Your actions will be logged (anonymously)

#### Thanks ...



for your attention!

