PROFESSUR
MEDIENINFORMATIK

01 Basic Concepts of Information Retrieval

Lecture Media Retrieval

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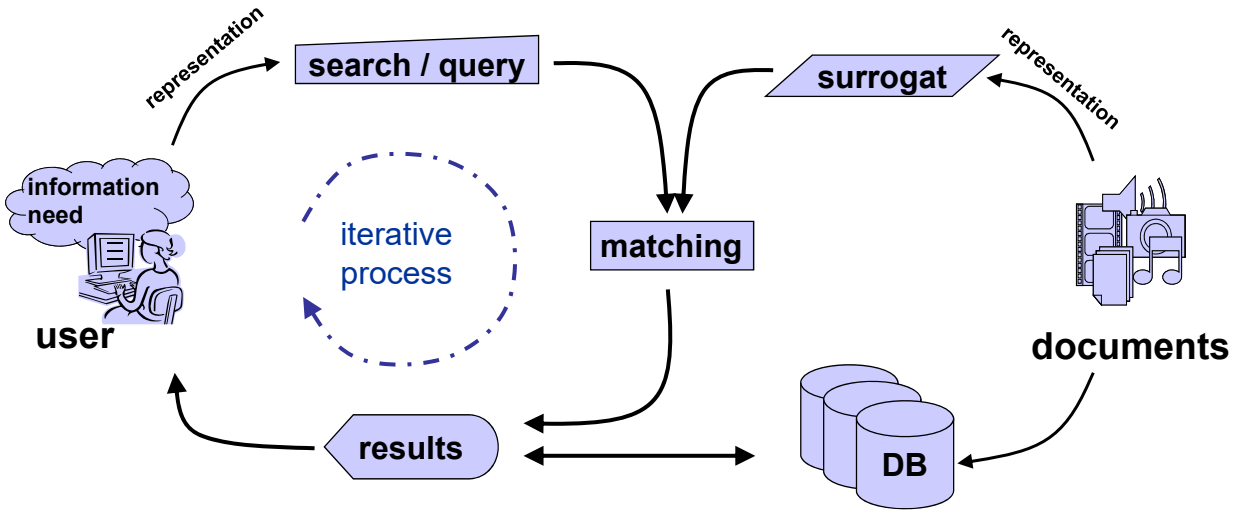
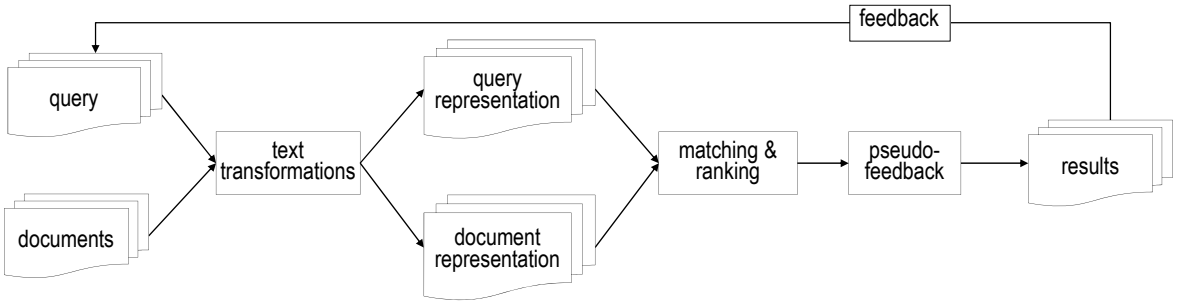
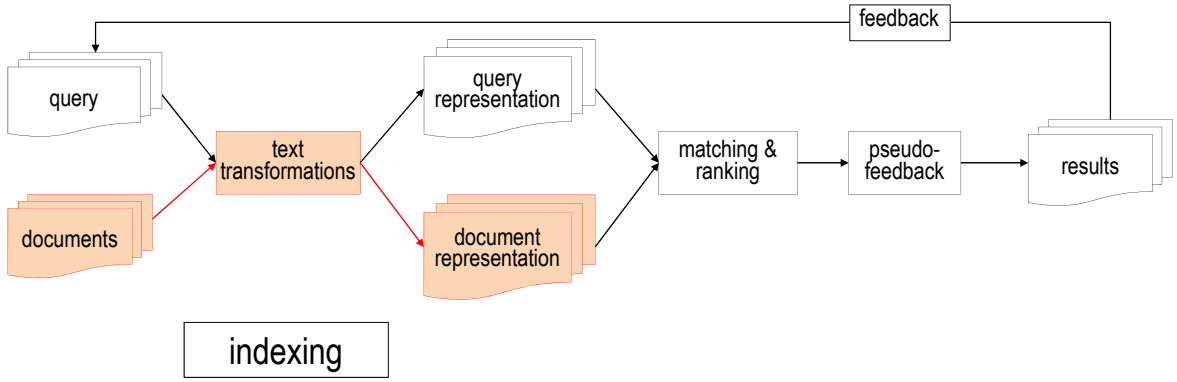
Information Retrieval in a Nutshell



IRS Components: Text Retrieval



IRS Components: Text Retrieval



Indexing (ANSI 1992)

The function of an index is to provide users with an efficient and systematic means for locating documents, portions of documents, or descriptions of documents that may address information needs or requests. An index should therefore:

- identify documents that treat particular topics or possess particular features;
- discriminate between major and minor treatments of particular topics or manifestations of particular features;
- provide access to topics or features by means of the terminology of users;
- link terms representing equivalent concepts and indicate relationships among terms representing related concepts;
- provide for the combination of terms to facilitate the identification of particular types or aspects of topics or features and to eliminate unwanted types or aspects.

Concepts of Indexing

- Manual / Intellectual Indexing
- Automatic Indexing
  - Statistical Indexing
  - Linguistical Indexing
- Semi-automatic Indexing
  - Relevance-Feedback
  - Computer Supported Indexing

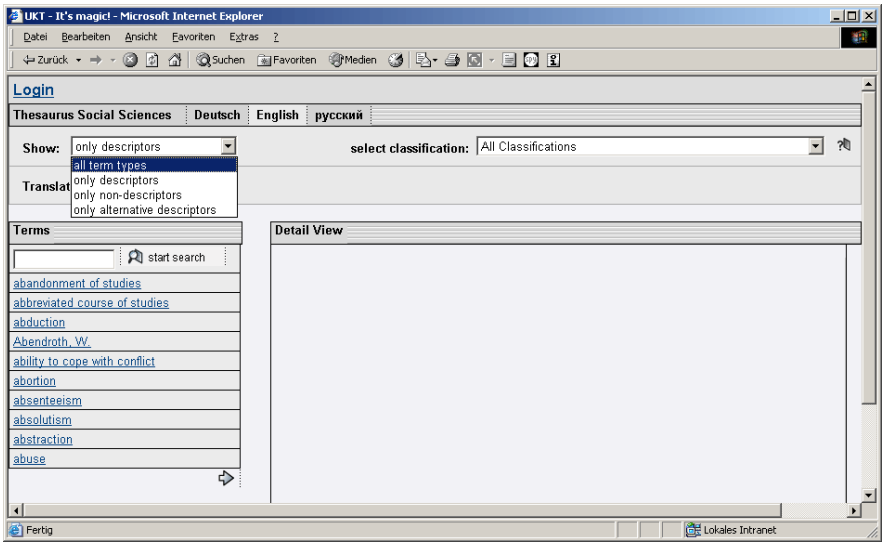
Manual / Intellectual Indexing

Thesaurus: Controlled Terms / Controlled Vocabulary

Thesaurus: Relations			
DIN 1463-1		ISO 2788	
BF	Benutzt für	UF	Used for
BS	Benutze Synonym	USE/SYN	Use synonym
OB	Oberbegriff	BT	Broader term
UB	Unterbegriff	NT	Narrower term
VB	Verwandter Begriff	RT	Related term
SB	Spitzenbegriff	TT	Top term

Example: http://sowiport.gesis.org/thesaurus

Manual / Intellectual Indexing: Thesaurus



Steps of Linguistic Indexing

- Free text

→Morphology

→Syntax

→Semantics

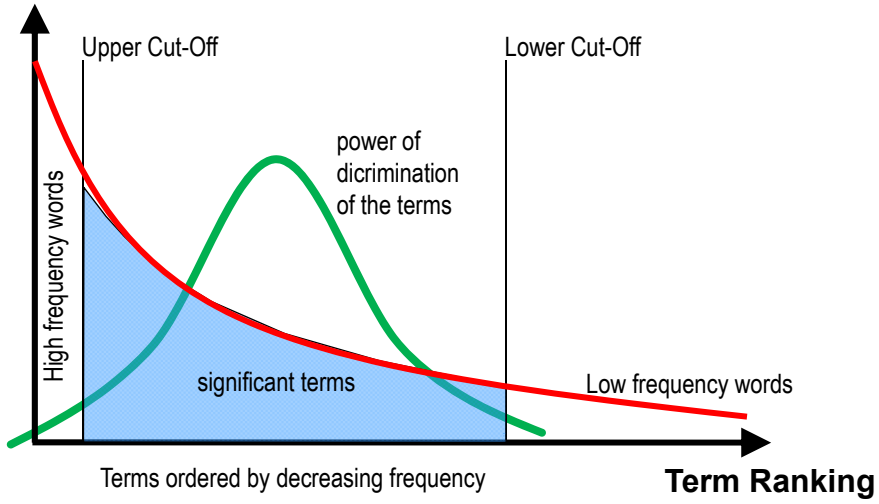
→Pragmatics

Statistical Features of Text

- word frequency
- ἅπαξ λεγόμενον (*hápax legómenon*)
- Zipf`s law (1949):
  frequency * ranking ≈ constant
- Luhn`s term weighting (1958)

Token	Appearance	%
the	7.398.934	5.9
of	3.893.790	3.1
to	3.320.687	2.7
and	3.320.687	2.6
in	2.311.785	1.8
is	1.559.147	1.2
for	1.313.561	1.0
The	1.144.860	0.9
that	1.066.503	0.8
said	1.027.713	0.8

Discriminating power of words



see:
Salton&McGill: Abb.3-2
Van Jijsbergen: Fig.2.1

Statistical Features of Text: How can we use it?

- Stop word list
  → High frequency words (is, this, he, she, it, be, and, or, not …)
  Reduction of data approx. 40% (Baeza-Yates&Ribeiro-Neto: 167) to 70% (Salton&McGill:66)
- Low frequency words
- Later: term weighting

https://commons.wikimedia.org/wiki/File:Sarah-Bernhardt_(Hamlet).jpg

Steps of Linguistic Indexing

- Free text
  - → Morphology
    - → Syntax
      - → Semantics
        - → Pragmatics

Decompounding (*Kompositazerlegung*):	lifecycle → life, cycle
Derivation:	building → build, builder
Lemmatization (*Lemmatisierung*):	built → build
Stemming (*Stammformreduktion*):	construction, constructing construct, structure → struct
Problems:	Homographs: Heteronyms → It was about time to present the present. Homonyms → can, flat, rose

Steps of Linguistic Indexing

- Free text
  - → Morphology
    - → Syntax
      - → Semantics
        - → Pragmatics

Adjektive – noun (*Adjektiv-Substantiv*):	red carpet
Genitiv relation:	House *of* Cards, Victoria*'s* Secret
Coordinational relation:	Peter, Paul, *and* Mary
Prepositions:	*at* the door, *in* the house, *on* the roof

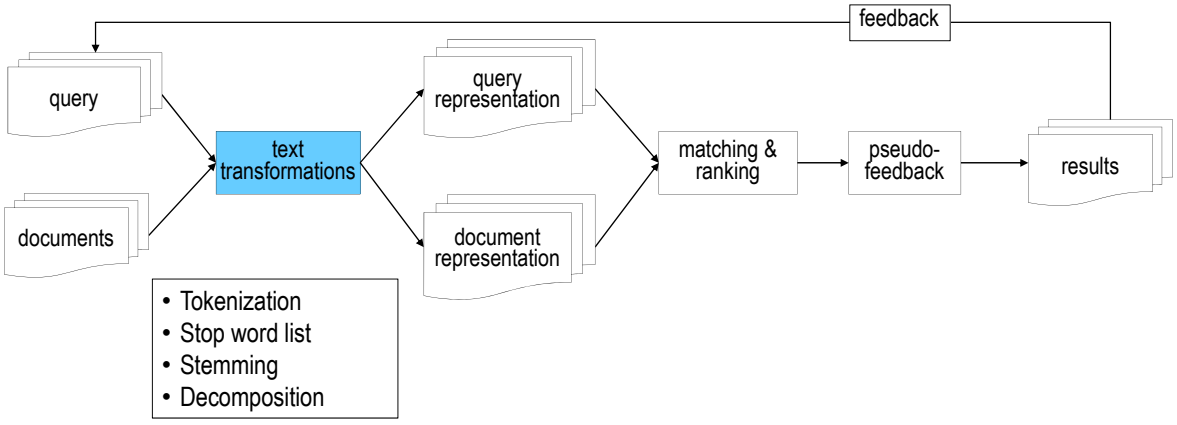
Steps of Linguistic Indexing

- Free text
  - → Morphology
    - → Syntax
      - → Semantics
        - → Pragmatics

Knowledge representation: Thesaurus, Classifikation

Problem: Polysemie

Transformations in Text Retrieval



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Tokenization: Concepts

- Word: delimited string of characters as can be found in texts
- Term: normalized word (upper / lower case, ending, …)
- Token: instance of a word / term in a text
- Type: class of a token

- Example: „To be or not to be“
  - Words: be, be, not, or, To, to
  - Terms: be, be, not, or, to, to
  - Token: words / terms
  - Type: be, not, or, to

Stemming (*Stammformreduktion*)

1. *dictionary-based stemmers*
2. *n-gram stemmers*
3. *affix stemmers*

} algorithmic stemmers

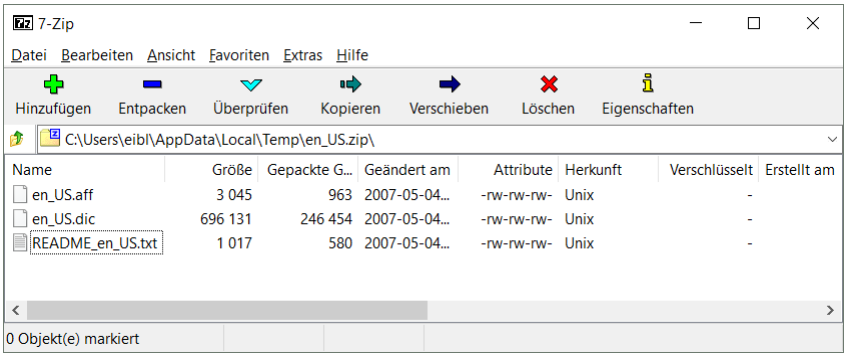
Stemming: affix

Definition by Merriam Webster: a letter or group of letters added to the beginning or end of a word to change its meaning.

a. prefix (*un*happy)
b. infix (German: ein*ge*schoben, ein*zu*schieben)
c. suffix (*suffix stripping*):
   - *a*-suffix, *attached* suffix (mandar*gli* = mandare + ***gli*** = to send + to him)
   - *i*-suffix, or *inflectional* suffix (fit + ***ed*** -> fitted (double ***t***))
   - *d*-suffix, or *derivational* suffix (medico + ***astro*** = medicastro = quack doctor, in English: -ness in some adjectives to get a noun)

Stemming I: dictionary-based stemmer

Hunspell Dictionary Stemmer



Stemming I: dictionary-based stemmer - Hunspell

Dictionary file:

```
3
hello
try/B
work/AB
```

Affix file:

```
SET UTF-8
TRY esianrtolcdugmphbyfvkwzESIANRTOLCDUGMPHBYFVKWZ'
REP 2
REP f ph
REP ph f
PFX A Y 1
PFX A 0 re .
SFX B Y 2
SFX B 0 ed [^y]
SFX B y ied y
```

→Accepted words with this dictionary and affix combination:
"hello", "try", "tried", "work", "worked", "rework", "reworked".

source: https://www.systutorials.com/docs/linux/man/4-hunspell/

Stemming II: n-gram stemmers

- *2-gram stemmer* (bigram)
- *3-gram stemmer* (trigram)
- *4-gram stemmer* (tetragram)

Example: HOUSE

Bigram → ▪H, HO, OU, US, SE, E▪

Trigram → ▪▪H, ▪HO, HOU, OUS, USE, SE▪, E▪▪

Tetragram → ▪▪▪H, ▪▪HO, ▪▪HOU, HOUS, OUSE, USE▪, SE▪▪, E▪▪▪

Stemming II: n-gram stemmers

T1: HOUSEKEEPING
T2: HOUSEHOLD
T3: SLAUGHTERHOUSE
T4: TROUSER

How similar are these terms?

(on a trigram-basis…)

Stemming II: n-gram stemmers

T1	T2	T3	T4
HOUSEKEEPING	HOUSEHOLD	SLAUGHTERHOUSE	TROUSER
▪▪H	▪▪H	▪▪S	▪▪T
▪HO	▪HO	▪SL	▪TR
HOU	HOU	SLA	TRO
OUS	OUS	LAU	ROU
USE	USE	AUG	OUS
SEK	SEH	UGH	USE
EKE	EHO	GHT	SER
KEE	HOL	HTE	ER▪
EEP	OLD	TER	R▪▪
EPI	LD▪	ERH	
PIN	D▪▪	RHO	
ING		HOU	
NG▪		OUS	
G▪▪		USE	
		SE▪	
		E▪▪	
14	11	16	9

T1∩T2	▪▪H	▪HO	HOU	OUS	USE
T1∩T3	HOU	OUS	USE		
T1∩T4	OUS	USE			
T2∩T3	HOU	OUS	USE		
T2∩T4	OUS	USE			
T3∩T4	OUS	USE			

→ Distance

Distance

Similarity measures: Dice, Jaccard, Cosinus, Ochiai, a.s.o.

Two very simple measures:

						Count equivalent: **Tx∩Ty**	Count equivalent in relation to length: $\frac{\mathbf{Tx \cap Ty}}{\mathbf{Tx+Ty}}$
T1∩T2	▪▪H	▪HO	HOU	OUS	USE	5	5 / (14+11) = 0,2
T1∩T3	HOU	OUS	USE			3	3 / (14+16) = 0,1
T1∩T4	OUS	USE				2	2 / (14+9) = 0,08

Stemming III: affix stemmers

classical affix-stemmers:

- Porter Stemmer
- Snowball Stemmer
- Krovetz (Kstem)

M.F. Porter: An Algorithm for suffix stripping

Terms with a common stem will usually have similar meanings, for example:

```
CONNECT
CONNECTED
CONNECTING
CONNECTION
CONNECTIONS
```

Porter Stemmer: Vowel-Consonant-Sequences

A consonant will be denoted by `c`, a vowel by `v`. A list `ccc...` of length greater than 0 will be denoted by `C`, and a list `vvv...` of length greater than 0 will be denoted by `V`. Any word, or part of a word, therefore has one of the four forms:

```
CVCV ... C
CVCV ... V
VCVC ... C
VCVC ... V
```

These may all be represented by the single form

```
[C]VCVC ... [V]
```

where the square brackets denote arbitrary presence of their contents. Using `(VC){m}` to denote `VC` repeated *m* times, this may again be written as

```
[C](VC){m}[V].
```

Porter Stemmer: VC - Vowel Consonant Sequences

m will be called the \measure\ of any word or word part when represented in this form. The case m = 0 covers the null word. Here are some examples:

```
m=0 TR, EE, TREE, Y, BY.
m=1 TROUBLE, OATS, TREES, IVY.
m=2 TROUBLES, PRIVATE,    OATEN,  ORRERY.
    ccvvccvc  ccvcvcv     vvcvc   vccvcv
    CVCVC     CVCVCV      VCVC    VCVCV
    [C]VCVC   [C]VCVC[V]  VCVC    VCVCV
```

Porter Stemmer: *suffix removal*

The \rules\ for removing a suffix will be given in the form

```
(condition) S1 -> S2
```

This means that if a word ends with the suffix S1, and the stem before S1 satisfies the given condition, S1 is replaced by S2. The condition is usually given in terms of m, e.g.

```
(m > 1) EMENT ->
```

Here S1 is `EMENT' and S2 is null. This would map REPLACEMENT to REPLAC, since REPLAC is a word part for which m = 2. The `condition' part may also contain the following:

```
*S - the stem ends with S (and similarly for the other letters).
*v* - the stem contains a vowel.
*d - the stem ends with a double consonant (e.g. -TT, -SS).
*o - the stem ends cvc, where the second c is not W, X or Y (e.g. -WIL, -HOP).
```

Porter Stemmer: *suffix removal*

Step 1a

SSES	-> SS	caresses	-> caress
IES	-> I	ponies	-> poni
		ties	-> ti
SS	-> SS	caress	-> caress
S	->	cats	-> cat

Step 1b

(m>0) EED	-> EE	feed	-> feed
		agreed	-> agree
(*v*) ED	->	plastered	-> plaster
		bled	-> bled
(*v*) ING	->	motoring	-> motor
		sing	-> sing

Porter Stemmer: *suffix removal*

If the second or third of the rules in Step 1b is successful, the following is done:

AT -> ATE	conflat(ed) -> conflate
BL -> BLE	troubl(ed) -> trouble
IZ -> IZE	siz(ed) -> size
(*d and not (*L or *S or *Z)) -> single letter	hopp(ing) -> hop
	tann(ed) -> tan
	fall(ing) -> fall
	hiss(ing) -> hiss
	fizz(ed) -> fizz
(m=1 and *o) -> E	fail(ing) -> fail
	fil(ing) -> file

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Porter Stemmer: *suffix removal*

Step 1c

(*v*) Y -> I	happy -> happi
	sky -> sky

Porter Stemmer: *suffix removal*

Step 2

(m>0) ATIONAL -> ATE	relational -> relate
(m>0) TIONAL -> TION	conditional -> condition
	rational -> rational
(m>0) ENCI -> ENCE	valenci -> valence
(m>0) ANCI -> ANCE	hesitanci -> hesitance
(m>0) IZER -> IZE	digitizer -> digitize
(m>0) ABLI -> ABLE	conformabli -> conformable
(m>0) ALLI -> AL	radicalli -> radical
(m>0) ENTLI -> ENT	differentli -> different
(m>0) ELI -> E	vileli - > vile
(m>0) OUSLI -> OUS	analogousli -> analogous
(m>0) IZATION -> IZE	vietnamization -> vietnamize
(m>0) ATION -> ATE	predication -> predicate
(m>0) ATOR -> ATE	operator -> operate
(m>0) ALISM -> AL	feudalism -> feudal
(m>0) IVENESS -> IVE	decisiveness -> decisive
(m>0) FULNESS -> FUL	hopefulness -> hopeful
(m>0) OUSNESS -> OUS	callousness -> callous
(m>0) ALITI -> AL	formaliti -> formal
(m>0) IVITI -> IVE	sensitiviti -> sensitive
(m>0) BILITI -> BLE	sensibiliti -> sensible

Porter Stemmer: *suffix removal*

Step 3

```
(m>0) ICATE ->  IC     triplicate     ->  triplic
(m>0) ATIVE ->         formative      ->  form
(m>0) ALIZE ->  AL     formalize      ->  formal
(m>0) ICITI ->  IC     electriciti    ->  electric
(m>0) ICAL  ->  IC     electrical     ->  electric
(m>0) FUL   ->         hopeful        ->  hope
(m>0) NESS  ->         goodness       ->  good
```

Porter Stemmer: *suffix removal*

Step 4

```
(m>1) AL     ->          revival        ->  reviv
(m>1) ANCE   ->          allowance      ->  allow
(m>1) ENCE   ->          inference      ->  infer
(m>1) ER     ->          airliner       ->  airlin
(m>1) IC     ->          gyroscopic     ->  gyroscop
(m>1) ABLE   ->          adjustable     ->  adjust
(m>1) IBLE   ->          defensible     ->  defens
(m>1) ANT    ->          irritant       ->  irrit
(m>1) EMENT  ->          replacement    ->  replac
(m>1) MENT   ->          adjustment     ->  adjust
(m>1) ENT    ->          dependent      ->  depend
(m>1 and (*S or *T)) ION ->
                         adoption       ->  adopt
(m>1) OU     ->          homologou      ->  homolog
(m>1) ISM    ->          communism      ->  commun
(m>1) ATE    ->          activate       ->  activ
(m>1) ITI    ->          angulariti     ->  angular
(m>1) OUS    ->          homologous     ->  homolog
(m>1) IVE    ->          effective      ->  effect
(m>1) IZE    ->          bowdlerize     ->  bowdler
```

Porter Stemmer: *suffix removal*

Step 5a

```
(m>1) E     ->        probate        ->  probat
                      rate           ->  rate
(m=1 and not *o) E -> cease          ->  ceas
```

Step 5b

```
(m > 1 and *d and *L) -> single letter
                      controll       ->  control
                      roll           ->  roll
```

Porter Stemmer: *suffix removal*

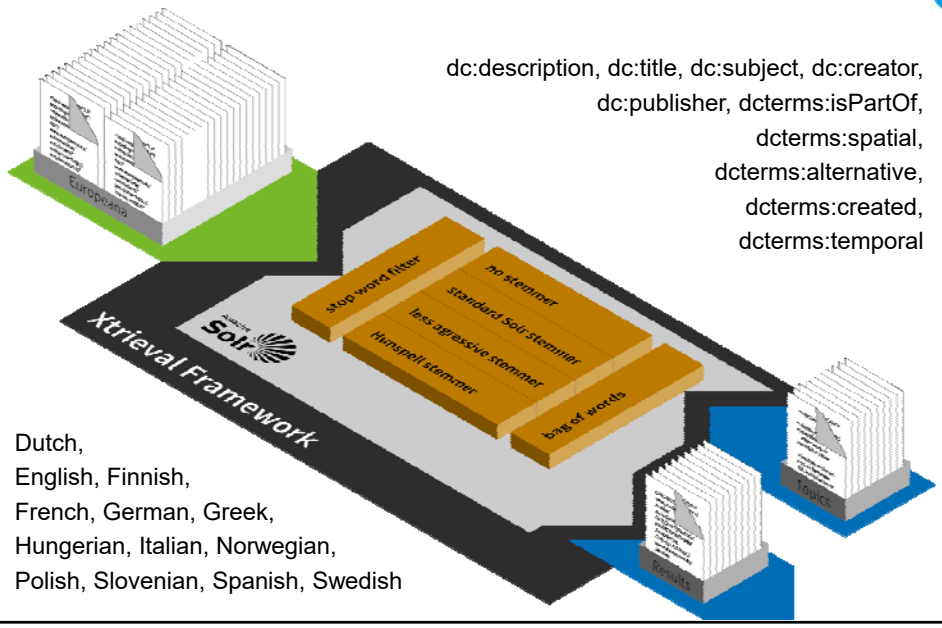
Complex suffixes are removed bit by bit in the different steps. Thus GENERALIZATIONS is stripped to GENERALIZATION (Step 1), then to GENERALIZE (Step 2), then to GENERAL (Step 3), and then to GENER (Step 4). […]

In a vocabulary of 10,000 words, the reduction in size of the stem was distributed among the steps as follows:

```
Suffix stripping of a vocabulary of 10,000 words
------------------------------------------------
Number of words reduced in step 1:   3597
         "                      2:   766
         "                      3:   327
         "                      4:   2424
         "                      5:   1373
Number of words not reduced:         3650
```

The resulting vocabulary of stems contained 6370 distinct entries. Thus the suffix stripping process reduced the size of the vocabulary by about one third.

Example: Cultural Heritage



dc:description, dc:title, dc:subject, dc:creator, dc:publisher, dcterms:isPartOf, dcterms:spatial, dcterms:alternative, dcterms:created, dcterms:temporal

Dutch, English, Finnish, French, German, Greek, Hungerian, Italian, Norwegian, Polish, Slovenian, Spanish, Swedish

Example Cultural Heritage

```
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<ims:fields>
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<dc:subject>Tachymarptis melba</dc:subject>
<dc:title>RundunZaqquBajda (Orn.0240)</dc:title>
<dc:title>Alpine Swift (Orn.0240)</dc:title>
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01A6A6C3BB655ACF9D</europeana:uri>
</ims:fields>
</ims:metadata>
```

Fig.1. Europeana CHiC Collection Sample Record

Vivien Petras, Toine Bogers, Nicola Ferro and Ivano Masiero (2013). Cultural Heritage in CLEF (CHiC) 2013 – Multilingual Task Overview: http://ceur-ws.org/Vol-1179/CLEF2013wn-CHiC-PetrasEt2013.pdf

Alpine Swift (*Rundun Zaqqu Bajda*)

Example Cultural Heritage

Table 5. Best Multilingual Experiments per Group (in MAP)

Participant	Experiment Identifier	Topic Languages	Collection Languages	MAP
Chemnitz	TUC_ALL_LA	All	All	23.38%
CEA List	MULTILINGUALNOEXPANSION	All NOT EL, HU, SL	All NOT EL, HU, SL	18.78%
Neuchatel	UNINEMULTIRUN5	All	All	15.45%
RSLIS	RSLIS_MULTI_FUSION_COMBSUM	All	All	8.37%
Westminster	R005	EN	EN,IT	6.30%
Berkeley	BERKMLENFRDE19	EN,FR,DE	EN,FR,DE	3.93%

Figure 3 shows the best 5 multilingual runs in an interpolated recall vs. average precision graph.

Vivien Petras, Toine Bogers, Nicola Ferro and Ivano Masiero (2013). Cultural Heritage in CLEF (CHiC) 2013 – Multilingual Task Overview: http://ims-sites.dei.unipd.it/documents/71612/430938/CLEF2013wn-CHiC-PetrasEt2013.pdf

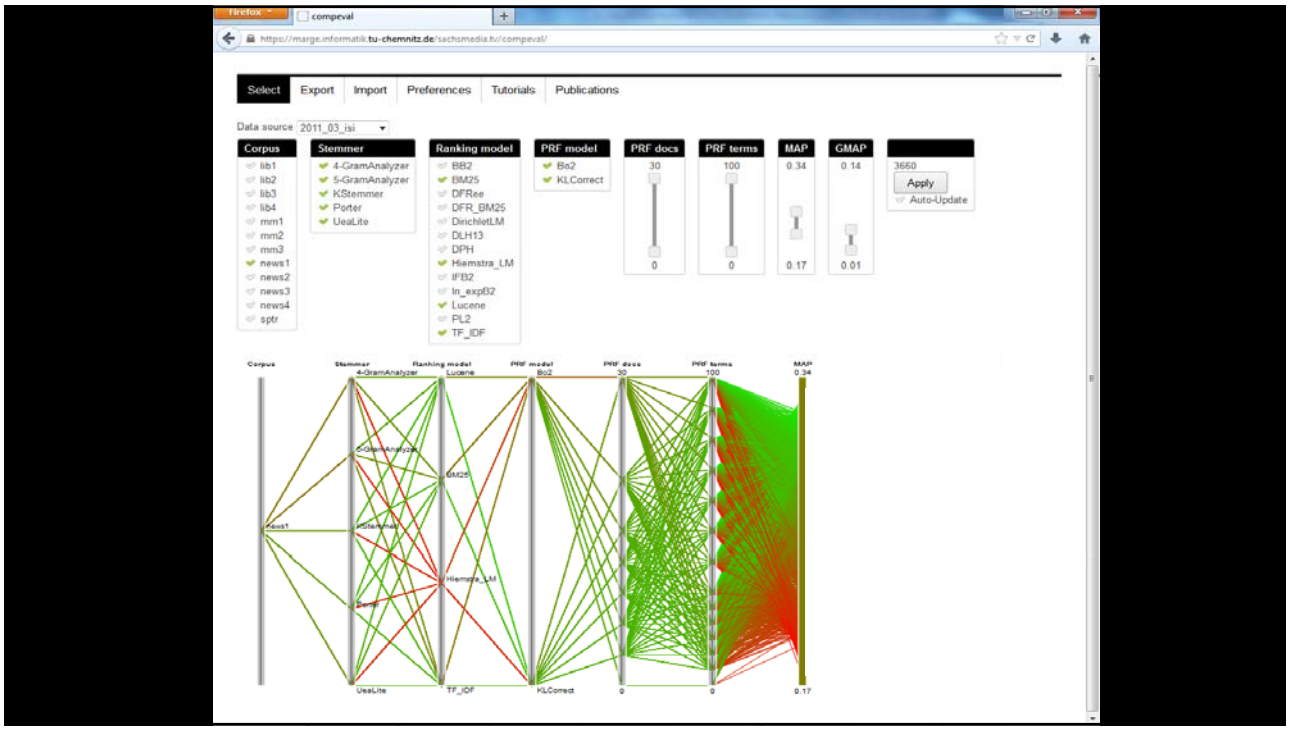
Example Cultural Heritage

Stemmer	MAP	GMAP	BPref	R-Precision
Solr standard	0.2583	0.1603	0.3538	0.3329
less aggressive	0.2590	0.1552	0.3686	0.3253
Hunspell	0.2466	0.1314	0.2914	0.3160
no stemmer	0.2684	0.1587	0.3031	0.3444
Snowball	0.2604	0.1591	0.3576	0.3360
no stop words extraction	0.1597	0.0621	0.2251	0.2297

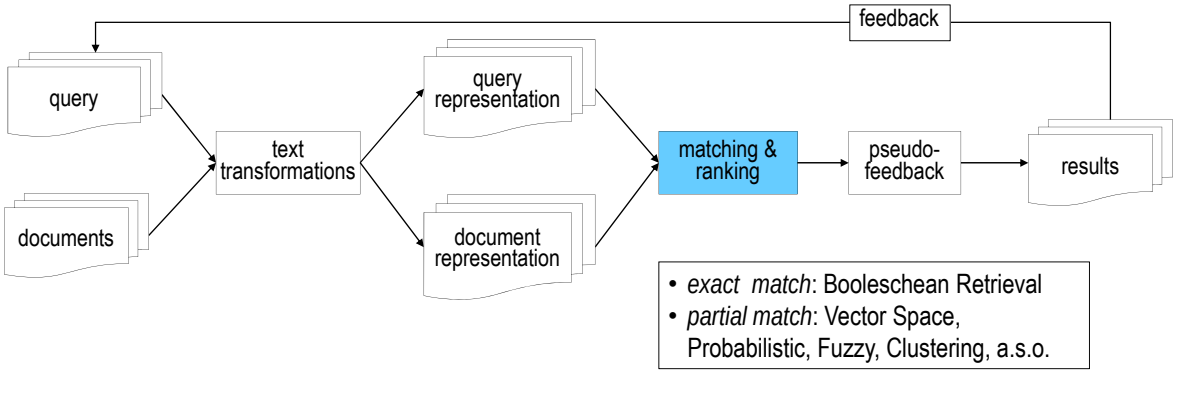
The higher the result the better (on a 0 to 1 range)

→ Obviously hughe impact of stop word list: Why?

→ Obviously negative impact of stemmer: Why?



Matching & Ranking



Information Retrieval Model

An information retrieval model is a quadrupel $[\mathbf{D}, \mathbf{Q}, \mathcal{F}, \mathcal{R}(q_i, d_j)]$ *where*

(1) $\mathbf{D}$ *is a set composed of logical views (or representations) for the documents in the collection.*

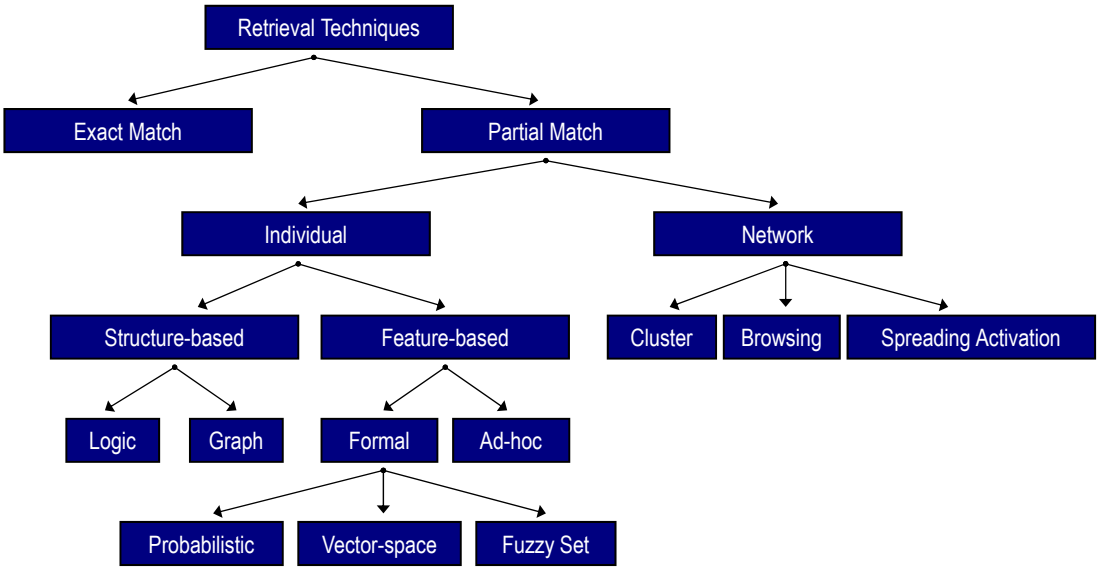
(2) $\mathbf{Q}$ *is a set composed of logical views (or representations) for the user information needs. Such representations are called queries.*

(3) $\mathcal{F}$ *is a framework for modeling document representations, queries, and their relationships.*

(4) $\mathcal{R}(q_i, d_j)$ *is a ranking function which associates a real number with a query* $q_i \in \mathbf{Q}$ *and a document representation* $d_j \in \mathbf{D}$*. Such ranking defines an ordering among the documents with regard to the query* q_i*.*

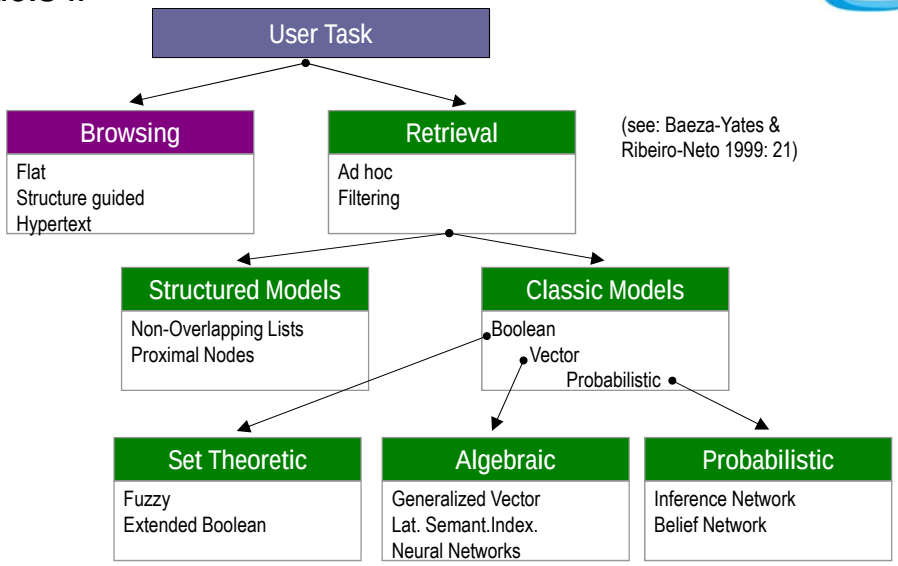
source: Baeza-Yates&Ribiero-Neto: 23

Retrieval Models I



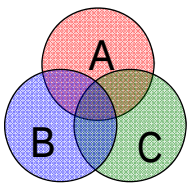
see: Belkin & Croft 1987: Fig.2

Retrieval Models II



Boolean Retrieval

- Boolean Logic
  - Conjunction: AND ∧
  - Disjunction 1: OR ∨ (Adjunction, not excluding)
  - Disjunction 2: XOR (Contravalence, excluding)
  - Negation: NOT ¬
- Binary weighting: $w_{i,j} \in \{0,1\}$



A	B	A∧B	A∧¬B	¬A∧B	¬A∧¬B	A∨B	A∨¬B	¬A∨B	¬A∨¬B	AXORB	AXOR¬B	¬AXORB	¬AXOR¬B
w	w	w	f	f	f	w	w	w	f	f	w	w	f
w	f	f	w	f	f	w	w	f	w	w	f	f	w
f	w	f	f	w	f	w	f	w	w	w	f	f	w
f	f	f	f	f	w	f	w	w	w	f	w	w	f

And now for something completely different …

Boolean Model – Example: How is Whisky made?

D1 Step 1 - Malting
Barley contains starch and it is this starch which needs to be converted into soluble sugars to make alcohol. For this to occur, the barley must undergo germination and this first part of the process is called 'malting'. Each distiller has their own preference about the type of barley they buy, but they need a type that produce high yields of soluble sugar. The barley is soaked for 2-3 days in warm water and then traditionally spread on the floor of a building called a malting house. It is turned regularly to maintain a constant temperature. This is also carried out on a commercial scale in large drums which rotate. [http://www.whiskyforeveryone.com/whisky_basics/how_is_whisky_made.html]

D2 1. Malting
Best quality barley is first steeped in water and then spread out on malting floors to germinate. It is turned regularly to prevent the build up of heat. Traditionally, this was done by tossing the barley into the air with wooden shovels in a malt barn adjacent to the kiln.
During this process enzymes are activated which convert the starch into sugar when mashing takes place. After 6 to 7 days of germination the barley, now called green malt, goes to the kiln for drying. This halts the germination. The heat is kept below 70°C so that the enzymes are not destroyed. Peat may be added to the fire to impart flavour from the smoke. [https://www.scotchwhiskyexperience.co.uk/about-whisky/making]

Boolean Model – Example: How is Whisky made?

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Boolean Model – Example: Document Representation

	alcohol	barley	enzymes	kiln	drum	malting	starch	sugar	temperature
D1	1	1	0	0	1	1	1	1	1
D2	0	1	1	1	0	1	1	1	0

Boolean Model – Example: Query

Query: bareley and (kiln or drum)

Representation: Disjunctive Normal Form DNF

→ Standardization of a logical formula: disjunction of conjunctive clauses like:

- A or B or C
- (A and B) or (A and C) or (B and C)

→ Usefull in automated processes

Boolean Model – Example: Query

Query: bareley and (kiln or drum)

Representation: Disjunctive Normal Form DNF

(bareley AND kiln) OR
(bareley AND drum) OR
(bareley AND kiln AND drum)

Boolean Model – Example: Document Representation

	alcohol	barley	enzymes	kiln	drum	malting	starch	sugar	temperature
D1	1	1	0	0	1	1	1	1	1
D2	0	1	1	1	0	1	1	1	0
(bareley AND kiln)		1		1	0				
V (bareley AND drum)		1		0	1				
V (bareley AND kiln AND drum)		1		1	1				

Boolean Model – Example: How is Whisky made?

D1 Step 1 - Malting
Barley contains starch and it is this starch which needs to be converted into soluble sugars to make alcohol. For this to occur, the barley must undergo germination and this first part of the process is called 'malting'. Each distiller has their own preference about the type of barley they buy, but they need a type that produce high yields of soluble sugar. The barley is soaked for 2-3 days in warm water and then traditionally spread on the floor of a building called a malting house. It is turned regularly to maintain a constant temperature. This is also carried out on a commercial scale in large drums which rotate.
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Boolean Model

- Advantages
  - Simple Model
  - Allows very specific queries
  - Easy to realize
- Disadvantages
  - Colloquial speech (*Umgangsprache*):
    - "AND" restricts / "and" expands
    - „or“ not exactly defined: excluding / including?
  - Negation
  - Nesting
  - No weighting: Binary relevance assessments are not natural

Ranking

- Presentation of result set (i.e. documents) in a specific order representing the relevance of the documents to the information need
- Problem: relevance assessment

Term Weighting I – tf

- Statistical assumption 1: The more often a term is used in a document, the better it describes the document.
- Examples:
  - D1: Apples are typical fruits of central Europe. In southern Europe you will find fruits which are more depending on sun light and mild climate like Oranges, Pineapples, …
  - D2: Apples are typical fruits of central Europe. There are many sorts of apples like the *Pink Lady*, a very sweet apple, or sour apples like *Granny Smith*. You can use apples in many ways. Just think about apple pie, apple juice, apple tea or apple crumble.
- Which document is about apples and which is about fruits in general?

Term Weighting I - tf

Term Frequency (tf)

$$f_{i,m} = \frac{freq_{i,m}}{\max_l freq_{l,m}}$$

with:

$freq_{i,m}$ = frequency of term t_i in document d_m

$\max_l freq_{l,m}$ = frequency of the term with the highest fequency in document d_m

$f_{i,m}$ = normalized frequency of term t_i in document d_m

Term Weighting II - idf

- Statistiscal assumption 2: The more specific a term is the less it is used.
- Example, Google search (Okt. 2013):

Term	Frequency
Farbe (color)	205.000.000
Grün (green)	27.000.000
Hellgrün (light green)	3.140.000
Birkengrün (birch green)	18.000
Frühlingsbirkengrün (spring birch green)	1.030

Term Weighting II - idf

Inverse Document Frequency (IDF): How specific is a term?

$$IDF_t = \log \frac{N - n_t}{n_t}$$

with:

t = term
n = documents related to t
N = number of documents in collection

→ Defines *term specifity* or *term exhaustivity* of a term in relation to a collection.
→ What do you need for a good query?

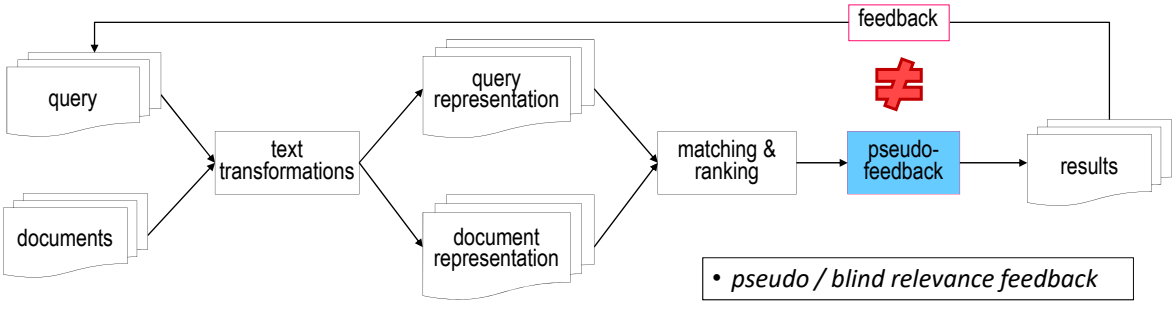
Summary Term Weighting

- Goal: High weighting for terms that appear quite often in only few documents.
- tf: term frequency factor (intra cluster similarity)
  - Frequency of term t_i in document d_m
  - Defines the ability of a term to describe a document
- idf: inverse document frequency (inter cluster independency)
  - Inverse frequency of a term t_i in the document collection
  - Terms used in many documents do not help discriminate relevant and non-relevant documents

(Baeza-Yates/Ribeiro-Neto, 1999,29)

Matching & Ranking

Finding: Top ranked documets are usually relevant and contain additional terms that are suitable for extending the query.



Pseudo Feedback:
Step 1: define a suitable amount X of top ranked documents
Step 2: extract a suitable amount Y of terms
Step 3: reformulate search query using these terms