Contents

Architecture Levels of IR Systems

Models for Interactive Retrieval

Information Seeking Behavior

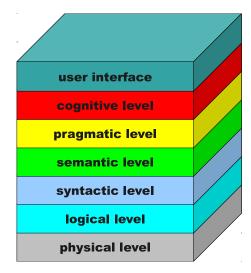
Information Searching

Strategic Support

From Cognitive Models to IR Interfaces

Summary

Architecture Levels of IR Systems



Interactive Information Retrieval

Norbert Fuhr

University of Duisburg-Essen

German IR Autumn School, Schloss Dagstuhl, 2012

UNIVERSITÄT DUISBURG ESSEN

Architecture Levels of IR Systems

- physical level: data structures and algorithms
- ► logical level: IR models
- syntactic level: focus on syntactic properties of objects (text as character sequence, image as pixel matrix, ...)
- \blacktriangleright semantic level: objects and their relationships \rightarrow ontologies, ...
- pragmatic level: meaning of a document/application issue \rightarrow *definition of relevance*
- **cognitive level**: user support during search *process*
- user interface

most systems comprise only some of these levels

- users have to compensate the missing levels (especially cognitive and pragmatic level)
- missing separation of levels
 (e.g. no logical or physical data independence as in DBMS)

Classical Probability Ranking Principle

defines optimum retrieval for probabilistic models: ranking documents according to decreasing values of the

probability of relevance

yields

optimum retrieval quality

Restrictions

- Relevance judgments of documents are independent
- ► Focus on user's assessment of result list

Models for Interactive Retrieval

A Probability Ranking Principle for Interactive IR Motivation Approach The Model Discussion Estimating IPRP Parameters via Gaze Tracking

- User has a rich set of interaction possibilities
 - (re)formulate query
 - selection based on summaries of various granularity
 - select related terms from list
 - follow document link
 - relevance judgment
- Information need changes during a search

No theoretic foundation for constructing IIR systems!

[Fuhr 08]

- Consider the complete interaction process
- ► Allow for different costs for different activities
- Allow for changes of the information need

Basic Assumptions

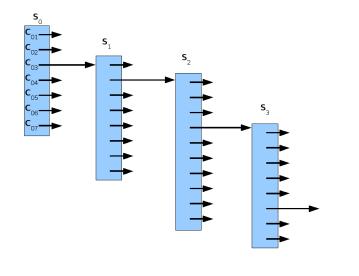
- Focus on a functional level of interaction (usability issues disregarded here)
- System presents list of choices to the user
- Users evaluate choices in linear order
- Only positive decisions/choices are of benefit for a user

Examples of decision lists

- ranked list of documents
- list of summaries
- list of document cluster
- KWIC list
- list of expansion terms
- links to related documents

▶ ...

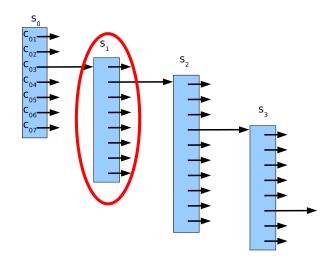
Abstraction: Situations with Lists of Choices



Basic ideas

- A user moves from situation to situation
- ► In each situation s_i, the user is presented a list of (binary) choices < c_{i1}, c_{i2},..., c_{i,ni} >
- The user decides about each of these choices sequentially
- \blacktriangleright The first positive decision moves the user to a new situation s_j
- ► A decision may be wrong, requiring backtracking

Probabilistic model focusing on single situation



Expected Benefit of a choice

- p_{ij} probability that the user will accept choice c_{ij}
- q_{ij} probability that this decision was right
- e_{ij} < 0: effort for evaluating the choice c_{ij}
- b_{ij} > 0: resulting benefit from positive, correct decision
- $g_{ij} \leq 0$: cost for correcting a wrong decision

Expected benefit of choice c_{ij}

$$E(c_{ij}) = e_{ij} + p_{ij} \left(q_{ij} b_{ij} + (1 - q_{ij}) g_{ij} \right)$$

Example

Web search: 'Java' \rightarrow *n*₀=290 mio. hits

System proposes extension terms:

term	ni	p_{ij}	b _{ij}	p _{ij} b _{ij}
program	195 mio	0.67	0.4	0.268
blend	5 mio	0.02	4.0	0.08
island	2 mio	0.01	4.9	0.049

benefit $b_{ij} = \log \frac{n_0}{n_i}$

Further remarks

$$E(c_{ij}) = e_{ij} + p_{ij} (q_{ij}b_{ij} + (1 - q_{ij})g_{ij})$$

- Expected benefit should be positive choices with negative values should not be presented to a user.
- Methods for estimating parameters p_{ij}, q_{ij}, b_{ij}, e_{ij}, g_{ij}: Issue of further research
- In the following, let a_{ij} = q_{ij}b_{ij} + (1 − q_{ij})g_{ij} ("average benefit")

$$E(c_{ij}) = e_{ij} + p_{ij}a_{ij}$$

Strategies for maximizing expected benefit

$$E(c_{ij})=e_{ij}+p_{ij}\left(q_{ij}b_{ij}+(1-q_{ij})g_{ij}
ight)$$

(assume that benefit b_{ij} and corr. effort g_{ij} are given)

- 1. minimize effort $|e_{ij}|$ but keep p_{ii} (selection prob.) and q_{ii} (success prob.) high
- 2. maximize p_{ij} : user should choose c_{ij} whenever it is appropriate

but keep success probability q_{ij} high \rightarrow increased effort e_{ii}

maximize q_{ij} by avoiding erroneous positive decisions
 → increased effort e_{ij}

Example for Expected Benefit

After formulating a query, a user may choose to perform the following actions with the corresponding parameter triple (e_{ij}, p_{ij}, a_{ij})

- 1. (-1.0, 0.3, 8) add expansion term to the query
- 2. (-2.0, 0.4, 10) look at the first result list entry
- 3. (-10.0, 0.4, 25) immediately go to the first document
- 4. (-5.0, 0.3, 20) look at an aggregated summary of the top ranking documents

In which order should these choices be presented to the user?

- 1. $(-1.0 + 0.3 \cdot 8) = 1.4$
- 2. $(-2.0 + 0.4 \cdot 10) = 2$
- 3. $(-10.0 + 0.4 \cdot 25) = 0$
- 4. $(-5.0 + 0.3 \cdot 20) = 1$

Expected benefit of a choice list

situation s_i with list of choices $r_i = \langle c_{i1}, c_{i2}, \dots, c_{i,n_i} \rangle$ expected benefit of choice list:

$$E(r_i) = e_{i1} + p_{i1}a_{i1} + (1 - p_{i1})(e_{i2} + p_{i2}a_{i2} + (1 - p_{i2})(e_{i3} + p_{i3}a_{i3} + \dots + (1 - p_{i,n-1})(e_{in} + p_{in}a_{in})))$$

$$= \sum_{j=1}^n \left(\prod_{k=1}^{j-1} (1 - p_{ik})\right)(e_{ij} + p_{ij}a_{ij})$$

Ranking of choices

Consider two subsequent choices c_{il} and $c_{i,l+1}$

$$E(r_i) = \sum_{\substack{j=1\\l\neq j\neq l+1}}^n \left(\prod_{k=1}^{j-1} (1-p_{ik})\right) (e_{ij} + p_{ij}a_{ij}) + t_i^{l,l+1}$$

where

$$t_{i}^{l,l+1} = (e_{il} + p_{il}a_{il})\prod_{k=1}^{l-1}(1 - p_{ik}) + (e_{i,l+1} + p_{i,l+1}a_{i,l+1})\prod_{k=1}^{l}(1 - p_{ik})$$

analogously $t_i^{l+1,l}$ for $<\ldots,c_{i,l+1},c_{il,},\ldots>$

Difference between alternative rankings

$$d_{i}^{l,l+1} = \frac{t_{i}^{l,l+1} - t_{i}^{l+1,l}}{\prod_{k=1}^{l-1} (1 - p_{ik})}$$

= $e_{il} + p_{il}a_{il} + (1 - p_{il})(e_{i,l+1} + p_{i,l+1}a_{i,l+1}) - (e_{i,l+1} + p_{i,l+1}a_{i,l+1} + (1 - p_{i,l+1})(e_{il} + p_{il}a_{il}))$
= $p_{i,l+1}(e_{il} + p_{il}a_{il}) - p_{il}(e_{i,l+1} + p_{i,l+1}a_{i,l+1})$

Necessary condition for the maximum expected benefit of the list: $d_i^{l,l+1} \stackrel{!}{\geq} 0$, which leads to

$$a_{il} + rac{e_{il}}{p_{il}} \geq a_{i,l+1} + rac{e_{i,l+1}}{p_{i,l+1}}$$

PRP for Interactive IR

$$a_{il} + rac{e_{il}}{p_{il}} \ge a_{i,l+1} + rac{e_{i,l+1}}{p_{i,l+1}}$$

~ Rank choices by decreasing values of

$$\varrho(c_{ij}) = a_{il} + \frac{\mathbf{e}_{il}}{p_{il}}$$

Expected Benefit vs. Ranking Criterion

expected benefit: $E(c_{ij}) = p_{ij}a_{ij} + e_{ij}$ ranking criterion: $\varrho(c_{ij}) = a_{il} + \frac{e_{il}}{p_{il}}$

choice	p _{ij}	a _{ij}	e _{ij}	E(c _{ij})	$\varrho(c_{ij})$
<i>c</i> ₁	0.5	10	-1	4	8
<i>c</i> ₂	0.25	16	-1	3	12

Expected benefits of the 2 possible lists:

 $E(\langle c_1, c_2 \rangle) = 4 + 0.5 \cdot 3 = 5.5$ $E(\langle c_2, c_1 \rangle) = 3 + 0.75 \cdot 4 = 6$

IIR-PRP: Observations

Rank choices by
$$a_{ij} + \frac{e_{ij}}{p_{ij}}$$

- *p_{ij}* 'probability of relevance' still involved
- ▶ tradeoff between effort *e_{ij}* and benefit *a_{ij}*
- difference between PRP and IIR-PRP due to variable values for *e_{ij}* and *a_{ij}*
- ▶ IIR-PRP looks only for the first positive decision

IIR-PRP vs. PRP

$$a_{il} + rac{e_{il}}{p_{il}} \ge a_{i,l+1} + rac{e_{i,l+1}}{p_{i,l+1}}$$

Assumptions for classical PRP:

- 1. constant effort for each document $e_{ij} = -E$, E > 0
- 2. constant benefit from each relevant document $a_{il} = B$

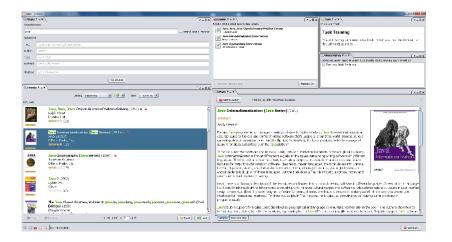
$$B - \frac{E}{p_{il}} \geq B - \frac{E}{p_{i,l+1}}$$
$$\Rightarrow p_{il} \geq p_{i,l+1}$$

 \rightsquigarrow Classic PRP still holds!

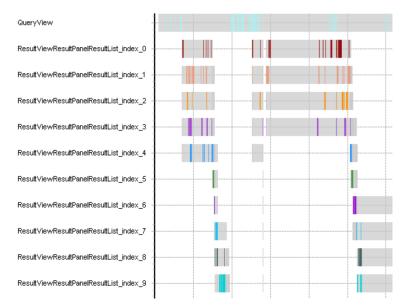
Parameter estimation

- Selection probability p_{ij}: focus of many IR models, but models for dynamic info needs required
- 2. Effort parameters e_{ij} , g_{ij} +success probability q_{ij} : most research needed
- 3. Benefit b_{ij} :
 - information value ?
 - saved effort

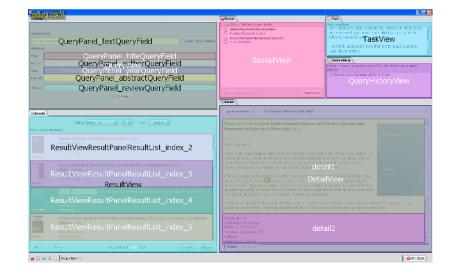
User Interface



AOI Sequence



Areas of Interest for Gaze Tracking

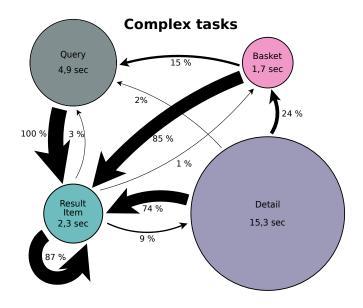


Task setting

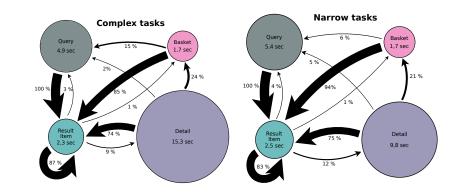
[Tran & Fuhr 12]

- ▶ Retrieval experiments with 12 subjects
- Users had to work on two tasks for 15 minutes each
 - complex tasks: consideration of user reviews necessary for judging relevance
 - narrow tasks: reading of book abstracts sufficient
- Users judge individually about relevance (by placing items in the basket

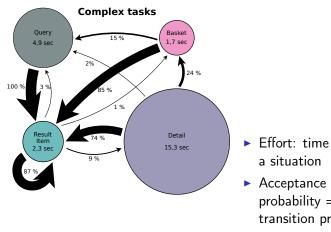
Markov Model for Complex Tasks



Markov Models Complex vs. Narrow Tasks



Estimating the iPRP parameters



- ► Effort: time spent in
- probability =transition probabilities
- ► Benefit?

Expected time for reaching the basket

- effort in states t_q , t_r , t_d and t_b
- p_{XY} : transition probability from state X to state Y
- expected times T_q , T_r and T_d for reaching the basket state

$$T_q = t_q + p_{qr}T_r$$

$$T_r = t_r + p_{rq}T_q + p_{rr}T_r + p_{rd}T_d$$

$$T_d = t_d + p_{dq}T_q + p_{dr}T_r$$

	complex	narrow
T_q	127.9	120.8
Tr	123.0	115.4
T_d	109.5	102.4
b_q	4.9	5.4
br	17.7	14.7
b _d	15.9	10.7

IPRP for Nonbinary Choices

- ► *c_{ij}*: choice
- m_{ij} alternatives c'_{iik} , $k = 1, \ldots, m_{ij}$
- ► *a*'_{*iik*} corresponding benefits
- q_{ijk} selection probabilities such that

$$\sum_{k=1}^{m_{ij}} q_{ijk} = 1$$
 and $\sum_{k=1}^{m_{ij}} q_{ijk} a'_{ijk} = a_{ij}$

Example: examining a result item in a complex task:

$$b_r = \frac{0.03(T_q - T_r) + 0.01(T_r - 0) + 0.09(T_d - T_r)}{0.03 + 0.01 + 0.09} = 17.7s$$

Models of information searching

classic IR

- content-oriented search in unstructured documents
- vague information needs, uncertain representations
- system-oriented view, assume static information need
- Interactive information retrieval
 - focus on user interaction with information system
 - dynamic information need
- ► 2 views on interactive IR:
 - 1. Information Seeking Behavior
 - 2. Information Searching

Information Seeking Behavior

Information Seeking Behavior and Information Searching Ellis' Behavioral Model of Information Seeking Strategies

Information Searching

- broader view than content-oriented search
- model user's actions, motivations and strategies for satisfying an information need
- questions of interest:
 - what triggers an information need?
 - what are users doing for solving this problem?

- focus on user's interaction with information sources
- regard classic IR systems as well as other sources (e.g. personal communication)

Ellis' Behavioral Model of Information Seeking Strategies

[Ellis 89]

- general model of search behavior
- based on empirical studies in social sciences and engineering companies
- general categories or properties of search behavior: Starting, Chaining, Browsing, Differentiating, Monitoring, Extracting, Verifying, Ending



Categories of search behavior according to Ellis (1)

Starting

- > get overview of literature/locate key authors in a field, e.g. by
 - selection of information source (e.g. personal collection, digital library, Web search engine)
 - review articles
 - personal contacts

Chaining

- follow different forms of referential connections between sources (in both directions)
 - citations
 - Web links
 - same author/ research team
 - ► same conference / journal issue
 - same category
- ► factors considered:
 - topical relevance
 - popularity of author
 - timeliness
 - citation frequency
 - cost and time for document acquisition
- leads to finding new sources or even to reformulation of information need

Categories of search behavior according to Ellis (3)

Differentiating

- judging of sources according to type, quality, importance, usefulness
- leads to information filtering
- ▶ e.g. comment vs. report, specification vs. manual

Categories of search behavior according to Ellis (3)

Browsing

- starts from information sources and retrieved documents
- semi-goal-oriented search by browsing in promising areas
- scanning of tables of contents, references, lists of people and organizations
- browsing is used when relevant information is available in a comprehensive way

Categories of search behavior according to Ellis (4)

Monitoring

- maintain awareness of developments and technologies in a field
- ▶ by following particular sources
 - formal channels: scientific journals, conferences, alert profiles
 - informal channels: personal contacts, actual practice (field research, experimental work)

Categories of search behavior according to Ellis (5)

Categories of search behavior according to Ellis (6)

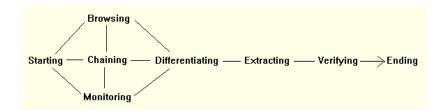
Extracting

- working through sources to locate material of interest
- ▶ material: documents, new sources, passages
- cognitive capture of information by the user
- user's background knowledge important

Verifying check information wrt. correctness and reliability

Ending end of search, linking of new information with previous knowledge

Process model



- no strict sequential process
- starting, browsing, chaining and monitoring are search procedures
- differentiating is a filtering step

Extension by Meho/Tibbo

- Repeated Ellis' study, new analysis (especially wrt. new technologies)
- ► confirmation of Ellis' model
- but: extension by new categories

Additional categories

Accessing

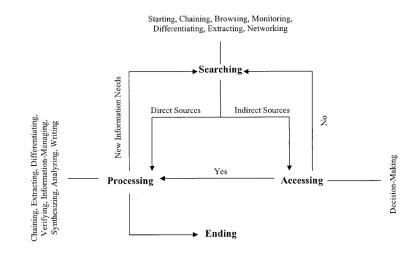
- Access to full texts (instead of surrogates)
- acquisition of contents via different channels and with different costs

Networking

- personal communication with various persons
- discussion and evaluation of retrieved information via internet/intranet fora

Information Managing

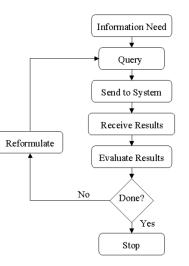
filing, storing and organizing retrieved and used information



Classical search process model

Information Searching

Simple models of the search process Patterns of search behavior Search activities Anomalous State of Knowledge Ingwersen's Cognitive Model Classification of search activities



Empirical studies

- information search consists of a sequence of connected, but different searches
- search result may trigger new searches
- only task context remains the same
- main goal of a search is accumulated learning and collection of new information while searching

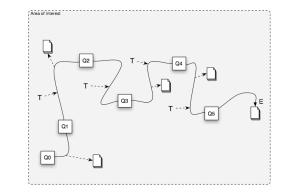
Support for Berrypicking

- ► Filing of single results
- Adding terms/items to the query

Berrypicking Model

[Bates 90]

- continuous change of information need and queries during search
- \blacktriangleright information need cannot be satisfied by a single result set
- instead: sequence of selections and collection of pieces of information during search

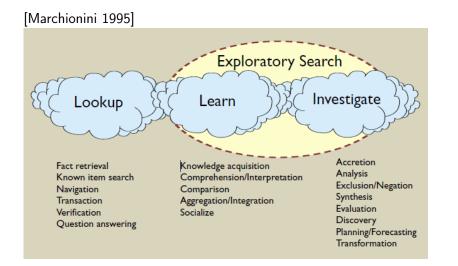


A taxonomy of Web search

[Broder 2002, Rose & Levinson 2004]
Navigational: to reach a particular site
Informational: to acquire information assumed to be present on one or more web pages
Transactional: to perform some web-mediated activity
Resource: to get access to an online resource

Search activities

Search modes





Search Modes: Lookup

Locate	Find a specific (possibly known) item
Verify	Confirm that an item meets some specific, objective criterion $% \left({{{\left[{{{C_{{\rm{B}}}}} \right]}_{{\rm{B}}}}} \right)$
Monitor	Maintain awareness of the status of an item for
	purposes of management or control

Search Modes: Learn

Compare	To identify similarities & differences between a set of items
Comprehend	To generate independent insight by understanding the patterns within a data set
Explore	To investigate an item or data set for the purpose of knowledge discovery

Search Modes: Investigate

- Analyze To examine an item or data set to identify patterns & relationships
- Evaluate To use judgement to determine the value of an item with respect to a specific goal
- Synthesize To create a novel or composite artefact from diverse inputs

[Belkin 80]

Classic IR systems: "best match" principle

- system returns those documents that fit best to the representation of the information need (e.g. query statement)
- only feasible, if user can give precise specification of her information need (like e.g. in DBMS)

Anomalous State of Knowledge (ASK)(2)

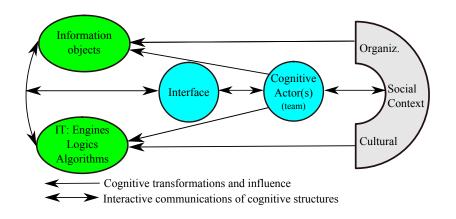
ASK-Hypothesis

- information need results from user's anomalous state of knowledge (ASK)
- user is unable to precisely specify information need for removing the ASK
- ► instead: describe ASK
- requires capture of cognitive and situation-specific aspects for resolving this anomaly

Ingwersen's Cognitive Model

- ► Global perspective
- comprises all factors influencing a search
 - social context
 - IR system
 - information objects
 - user interface
 - user
- focuses on *cognitive structures* manifestations of human cognition, reflexions and ideas

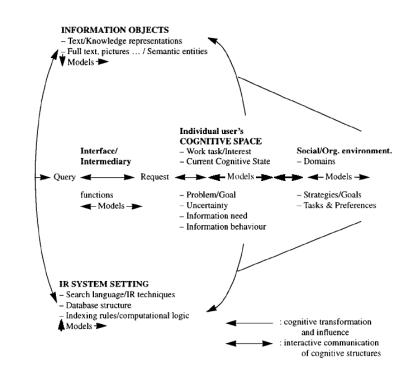
Ingwersen's Cognitive Model



Classification of search activities

[Cool & Belkin 2002]

- Access:
 - method: scanning ... searching
 - ▶ mode: recognition ... specification
- object interacted with
 - ► level: information ... meta-information
 - ▶ media: text, images, speech, video, ...
 - quantity: 1 object, set of objects, database
- common dimensions of interaction
 - ▶ information objects: parts ... complete objects
 - ► systematics: random ... systematic
 - degree: selective ... exhaustive
- interaction criteria
 - (e.g. precision, authority, date, person)



Polyrepresentation

[Ingwersen 94]

- representation of information objects in different forms
- \blacktriangleright representations should correlate with cognitive structures
- example: document can be represented by
 - title (specified by the author)
 - keywords (by indexer)
 - other documents citing the current doc (extern)
 - annotations (extern)
- retrieval system should supports several representations (thus, also several cognitive structures)
 - ightarrow intentional redundancy
- good search result, when several representations point to the same document (Overlap)

Polyrepresentation: the Amazon case

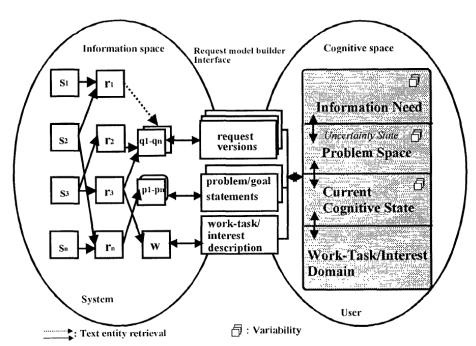


Polyrepresentation of the cognitive user space

cognitive space can be represented via polyrepresentation

- requests
- problems/goals
- ► work task

Global Polyrepresentation Model



Strategic Support

Information Seeking Behavior & Information Searching Levels of search activities Degrees of system involvement Proactivity in IR Systems

Information Seeking Behavior & Information Searching

- searching consists of sequence of different phases
- experienced searchers employ a variety of actions in different phases
- these actions should be supported by the system as much as possible
- each phase should be supported appropriately by the system

Bates' model for strategic system support

- Levels of search activities
- Degrees of system involvement

Monitoring Techniques

- CHECK To review the original request and compare it to the current search topic to see that it is the same.
- WEIGH To make a cost-benefit assessment, at one or more points of the search, of current or anticipated actions.
- PATTERN To make oneself aware of a search pattern, examine it, and redesign it if not maximally efficient or if out of date
- CORRECT To watch for and correct spelling and factual errors in one's search topic.
- RECORD To keep track of followed and of desirable trails not followed or not completed.

Levels of search activity

- 1. Move: An identifiable thought or *action* that is a part of information searching.
- 2. Tactic: One or a handful of moves made to *further* a search.
- 3. Stratagem: A larger, more complex set of thoughts and/or actions than the tactic, all designed to exploit the file structure of a *particular search domain* thought to contain desired information.
- 4. Strategy: A *plan*, which may contain moves, tactics, and/or stratagems, for an entire information search.

File Structure Tactics

- BIBBLE To look for a bibliography already prepared, before launching oneself into the effect of preparing one; more generally, to check to see if the search work one plans has already been done in a usable form by someone else.
- SELECT To break down complex search queries into subproblems and work on one problem at a time.
- SURVEY To review, at each decision point of the search, the available options before selection.
- CUT When selecting among several ways to search a given query, to choose the option that cuts out the largest part of the search domain at once .
- STRETCH To use a source for other than is intended purposes.
- SCAFFOLD To design an auxiliary, indirect route through the information files and resources to reach the desired information.
- CLEAVE To employ binary searching in locating an item in an ordered file.

Search Formulation Tactics

- SPECIFY To search on terms that are as specific as the information desired
- EXHAUST To include most or all elements of the query in the initial search formulation; to add one or more of the query elements to an already-prepared search formulation.
- REDUCE To minimize the number of the elements of the query in the initial search formulation; to subtract one or more of the query elements from an already-prepared search formulation.
- PARALLEL To make the search formulation broad (or broader) by including synonyms or otherwise conceptually parallel terms.
- PINPOINT To make the search formulation precise by minimizing (or reducing) the number of parallel terms, retaining the more perfectly descriptive terms.
- BLOCK To reject, in the search formulation, items containing or indexed by certain term(s), even if it means losing some document sections of relevance

Idea Tactics

- RESCUE In an otherwise unproductive approach, to check for possible productive paths still untried.
- BREACH To breach the boundaries of one's region of search, to revise one's concept of the limits of the intellectual or physical territory in which one searches to respond to a query.
- FOCUS To look at the query more narrowly, in one or both of two senses: (1) to move from the whole query to a part of it or (2) to move from a broader to a narrower conceptualization of the query.

Term Tactics

SUPER	To move upward hierarchically to a broader (superordinate) term.
SUB	To move downward hierarchically to a more specific (subordinate) term.
RELATE	To move sideways hierarchically to a coordinate term.
TRACE	To examine information already found in the search in order to find additional terms to be used in furthering the search.
VARY	To alter or substitute one's search terms in any of several ways. See remaining term tactics for some specific variations.
REARRANGE	To reverse or rearrange the words in search terms in any or reasonable orders.
CONTRARY	To search for the term logically opposite that describing the desired information.
RESPELL	To search under a different spelling.
RESPACE	To try spacing variants
NEIGHBOR	To seek additional search terms by looking at neighboring terms, whether proximate alphabetically, by subject similarity, or otherwise.
FIX	To try alternative affixes, whether prefixes, suffixes, or infixes.

Example stratagems

Subject Search

- Journal Run Having identified a journal central to one's topic of interest, one reads or browses through issues or volumes of the journal.
- Citation Search Using a citation index or database, one starts with a citation and determines what other works have cited it.
- Area Scan After locating a subject area of interest in a classification scheme, one browses materials in the same general area.
- Footnote Chase One follows up footnotes or references, moving backward in time to other related materials.

Tactics suggested in response to searcher request

SEARCHER COMMAND	SYSTEM RESPONSE LIST
Too many hits	SPECIFY EXHAUST PINPOINT BLOCK SUB
Too few hits	NEIGHBOR TRACE PARALLEL FIX SUPER RELATE VARY

SEARCHER COMMAND	SYSTEM RESPONSE LIST
No hits	RESPACE
	RESPELL
	REARRANGE
	CONTRARY
	SUPER
	RELATE
	NEIGHBOR
	TRACE
Need other terms	NEIGHBOR
or	TRACE
wrong terms	SUPER
	SUB
	RELATE

SEARCHER COMMAND	SYSTEM RESPONSE LIST
Revise terms	SPACE RESPELL FIX REVERSE CONTRARY SUPER SUB RELATE
Revise search formulation	SPECIFY EXHAUST REDUCE PARALLEL PINPOINT BLOCK

Degrees of system involvement

0 – No system involvement All search activities human generated and executed.
 Displays possible activities. System lists search activities when asked. Said activities may or may not also be executable by system (higher levels).
2 — Executes activities on command System executes specific actions at human command.
3 — Monitors search and recommends System monitors search process and recommends search activities:
a) Only when searcher asks for suggestions.b) Always when it identifies a need.
4 — Executes automatically. System executes actions automatically and then:
a) Informs the searcher.b) Does not inform the searcher.

Monitors search and recommends

Executes automatically + informs

AHOO!	summer school		Search
	summer school movie summer school movie summer school programs harvard summer school online summer school summer school activities summer school in london uela summer school summer school summer school	SUMMER SCHOOL, (2009) Unrated After a marathon horror-movie session, a teenage film	Summer School

Google	summerscool nijmegen About 399,000 results (0.12 seconds)
Q Everything	Showing results for summer school nijmegen. Search instead for summer scool nijmegen.
🙆 Images	Places for summer school near Nijmegen, Netherlands
🞬 Videos	Stg. European Summer School for Whole Animal Pharmacology ♀ Place page maps.google.com - Franciscanenstraat 16, Arnhem
🛗 News	
🕣 Shopping	Utrecht Summer School Q - Place page www.utrechtsummerschool.nl - Campusplein 7-8, Utrecht - 030 2534400
Places	Op Maat studiebegeleiding Q, - Flace page www.opmaatstudiebegeleiding.nl - Koningsweg 61, 's-Hertogenbosch - 073 6894386
More	 www.opmaatstudebegereiding.in - Koningsweg of, s-hertogenbosch - 0/3 0094300
Show search tools	International Summer School "Nijmegen03" ৰো Q nijmegen03.hef.kun.nl/- Cached The International Summer School on Particle and Nuclear Astrophysics (NUMEGEN03) is intended for graduate students and young postdocs in nuclear physics,

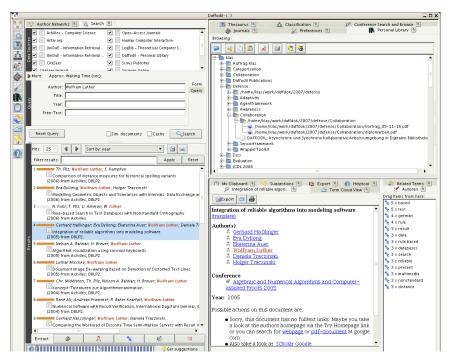
Executes automatically + does not inform

Coorla	
Google	summer school nimwegen
-	About 164,000 results (0.33 seconds)
Q Everything	Places for summer school near Nijmegen, Netherlands
🖸 Images	Stg. European Summer School for Whole Animal Pharmacology Q - Place page maps.google.com - Franciscanen straat 16, Arnhem
🞬 Videos	Utrecht Summer School Q - Place page www.utrechtsummerschool nl - Campusplein 7-8, Utrecht - 030 2534400
News Shopping	Op Maat studiebegeleiding - Place page www.opmaatstudiebegeleiding.nl - Koningsweg 61, 's-Hertogenbosch - 073 6894388
Places	www.opinaatatuurebegeletuing.in - Kuningsweg 01, shietugelebusti - 070 0034000
More	International Summer School "NJMEGEN09" ன 🔍 nijmegen09.hef.kun.nl/- Cached
Show search tools	The 3rd International Summer School on Astroparticle Physics (NIJMEGEND9) is intended for graduate students and postdocs working (or starting to work) in

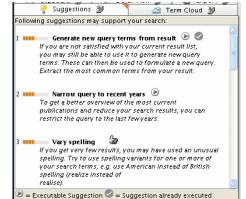
Combination of search activities and system support

Search Activity/ System involvement	Move	Tactics	Stratagem	Strategy
No system involvement				
Displays possible activities				
Executes activities on command				
Monitors search and recommends				
Executes automatically				

Daffodil desktop



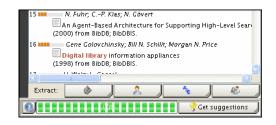
Daffodil: Search Continuation 2



- suggestions displayed as ranked list
- descriptive title, explanation, success rate
- execute on or more suggestions, with following feedback
- icons indicate the state of suggestions: (executable, used, useful)

Daffodil: Search Continuation

- proposal based on automatic analysis of the curent search result
- case-based reasoning
- availability of suggestions indicated as button at the bottom of result list window



Evaluation of search suggestions

- > 24 test subjects, half of them w/ suggestion component
- each subject worked on 3 tasks
- case base contained 30 different suggestions

Results:

supported users

- are more content with the search process (p = 0, 067)
- ▶ are significantly more satisfied with the result
- find more relevant documents
- use significantly more often Daffodil's advanced seearch tools

(unsupported users mainly restrict on reformulating queries)

Session Support

From Cognitive Models to IR Interfaces

Session support in the user interface Design Patterns for Search Modes Support for seeking behavior according to Ellis/Meho/Tibo

- Show results together with the query
- Allow editing of the previous query
- Show search history
- Allow for combination of queries
- Filing of single results
- Storing of sessions

Show results together with the query

as HyREX query: wsu	m(1.0,/#PCDATA \$title:stemen\$ "retrieval",1.0,/#PCDATA \$title:stemen\$ "in",1.0,/#PCDATA \$title:stemen\$ "o	ontext")
Results		
00 documents found,	100 documents displayed (with PIRE)	
Massimo Melucci (200 A Basis for Info	 ation Retrieval in Context. ACM Transactions on Information Systems 26(3) 	
	I C. Richardson; Thomas R. Roth-Berghofer; Laure Vieu (eds.) (2007). ing Context 6th International and Interdisciplinary Conference (CONTEXT). Springer, Berlin et al	
	io Di Marco; Davide Salvi (1998). y Context. Journal of Universal Computer Science 4(9)	

Allow editing of the previous query



Web Ergebnisse 1 - 10 von ungefähr 19 für uni-colleg duisburg-essen 2005. (0,58 Sekunden)

Uni-Colleg Duisburg

... Das Duisburger **Uni-Colleg**, ein Mix aus allgemein verständlichen Vorträgen, ... Auch im Wintersemester 2004 / **2005** freuen sich die Colleg-Organisatoren ... www.uni-duisburg-essen.de/ presse/events/uni-colleg.shtml - 9k - <u>Im Cache</u> - <u>Ahnliche Seiten</u>

Veranstaltungen des Uni-Colleg - Sommersemester 2005 ... Das Duisburger Uni-Colleg, ein Mix aus allgemein verständlichen Vorträgen, ... April 2005 © Universität Duisburg-Essen - Kontakt: ... www.uni-duisburg-essen.de/ presse/events/uni_colleg_sommer2005.shtml - 11k -Im Cache - Ähnliche Seiten [Weitere Ergebnisse von www.uni-duisburg-essen.de]

Show search history

PubMed Advanced Search

Search Box	Limits Details Help Search Preview Clear	Pub U.S. National Libr National Institutes
Search Builder		Display Settings: (
All Fields	AND C Add to Search B	Results: 1 to In vivo post Lim S, Kim Biotechnol Pr PMID: 21732 Engineerin
Search History		 Kim YM, Ch Biotechnol Bi
Search ₩5 Search escherichia coli treatment ₩4 Search escherichia coli symptoms ₩3 Search escherichia coli ₩2 Search Escheria coli ₩2 Search Escheria coli ₩2 Search escheria coli ₩2 Search Escheria coli ₩1 Search ehec virus	Most Recent Queries	PMID: 21732

Filing of single results

Dis	lay Settings. 🖂 Summary, 20 per page, Sorted by Recently Added	Send to: 🕑	Clipboard: <u>7 items</u>
Cli	pboard: 7 Remove all items		Filter your results:
1.	In vivo post-translational modifications of recombinant mussel adhesive protein in insect cells. Lim S, Kim KR, Choi YS, Kim DK, Hwang D, Cha HJ. Biotechnol Prog. 2011 Jun 16. doi: 10.1002/dtpr.662. [Epub ahead of print] PMID: 21732552 [PubMed - as supplied by publisher] Remove from clipboard		All (7) Free Full Text (0) Review (0)
2.	Engineering the pertose phosphate pathway to improve hydrogen yield in recombinant Escheric Kim YM, Cho HS, Jung GY, Park JM. Biotechnol Bioeng. 2011 Jul 5. doi: 10.1002/bit.23259. [Epub ahead of print] PMID: 21732330 [PubMed - as supplied by publisher] Remove from clipboard	<u>chia coli.</u>	Save Results in Co
□ 3.	Cell surface display of carbonic anhydrase on Escherichia coli using ice nucleation protein for sequestration.	<u>CO(2)</u>	Ma Interpreter Deter Chelmen annumbles server annumbles server deta samplation, nervere deta samplation, nervere detascubes silves annumbles

Allow for combination of queries

P	ub Med.gov	Search: PubMed	\$	RSS RSS	Save search Limits	
	. National Library of Medicine ional Institutes of Health	#1 or #3				Search
Disp	lay Settings: ⊙ Summary, 20 per pa	e, Sorted by Recently Added			Send to: 🗸	Filter your r
						All (2834)
Re	sults: 1 to 20 of 283422			Page 1 of 1417:	2 Next > Last >>	Free Full
	In vivo post-translational modifi	ations of recombinant mu	issel adhesive prot	tein in insect cells	<u>.</u>	Review (1
1.	Lim S, Kim KR, Choi YS, Kim Dł	, Hwang D, Cha HJ.				
	Biotechnol Prog. 2011 Jun 16. doi: 10	.1002/btpr.662. [Epub ahead o	f print]			
	PMID: 21732552 [PubMed - as suppl	ed by publisher]				
_						Save Resu
	Engineering the pentose phosp		nydrogen yreid in re	ecombinant Esche	erichia coli.	PS NOR Resources 11 In
2.	Kim YM, Cho HS, Jung GY, Park	JM.				My NCBI —
	Biotechnol Bioeng. 2011 Jul 5. doi: 1).1002/bit.23259. [Epub ahead	of print]			Select AL None 8
	PMID: 21732330 [PubMed - as suppl	ed by publisher]				C My Balkagenutry

Save sessions for continuing later

Time	Database	Туре	Term
Yesterday 1:59 PM	PubMed	record	In vivo post-translational modifications of recombinant
Yesterday 1:48 PM	PubMed	search	(ehec virus) OR (escherichia coli)
Yesterday 1:43 PM	PubMed	search	escherichia coli treatment
Yesterday 1:42 PM	PubMed	search	escherichia coli symptoms
Yesterday 1:41 PM	PubMed	search	Escheria coli
Yesterday 1:36 PM	PubMed	search	ehec virus
			<u>Clear</u> <u>Turn Off</u>

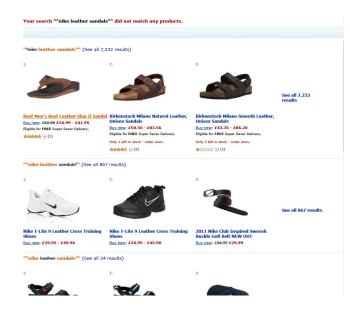
Design Patterns for Search Modes

Support for Marchionini's search modes

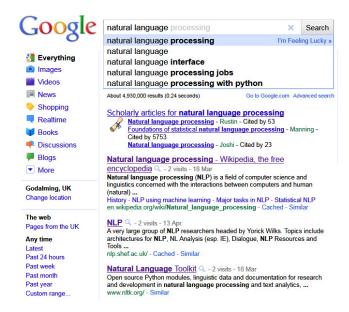
Locating: Autocomplete



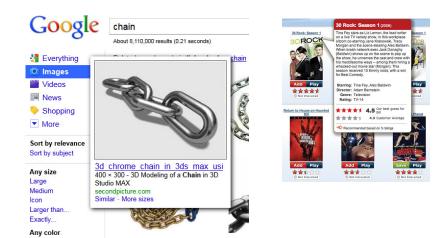
Locating: Partial Matches



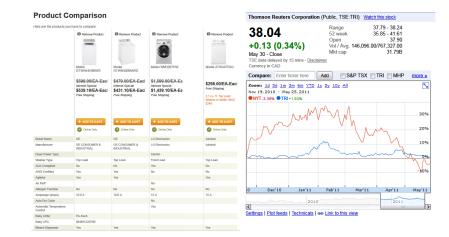
Verifying: Instant Results



Verifying: Detail Overlay



Comparing: Parallel Views

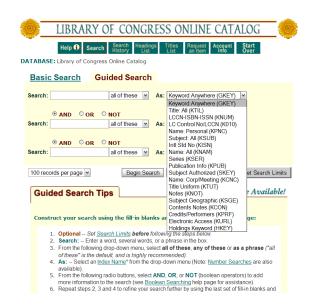


Exploring: Autosuggest

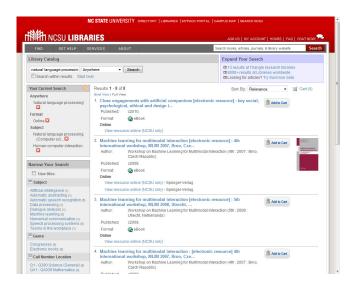


co.uk	n or register
golf	
golf golf clubs vw golf golf gti	T DE
mk2 golf mk1 golf golf balls golf bags	208
golf r32 vw golf mk4	25
Hide eBay suggestions 🧏	1 1

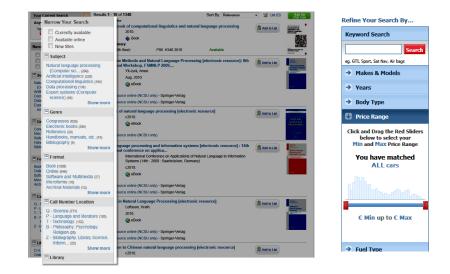
Exploring: faceted search



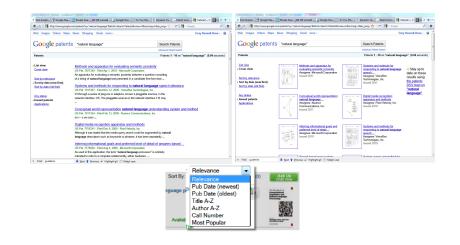
Exploring: faceted search



Comprehending: facet menus



Analyzing: alternate views



Analyzing: data visualization



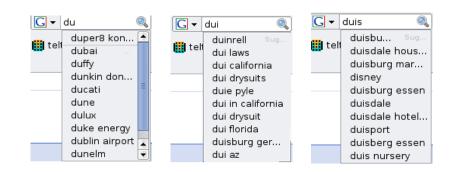
Support for seeking behavior according to Ellis/Meho/Tibo

Starting	Resource selection identifying popular authors entering search terms
Browsing	Sort result list by different criteria
	highlighting, also user-defined
Chaining	links in results pages
	comparison of result pages
Monitoring	storing and periodical execution of queries
Extracting	searching in the result page
Inform. Mgmt.	Collate/organize result items
	Annotate items (Interpret)

Starting: Resource Selection

CRefresh	ce ? _ All	Enter filter terms
🗹 ACM DL	٢	The digital library of the ACM - Association for Computing Machinery
DBLP	٦	The DBLP Computer Science Bibliography.
🗹 LEABIB	тип	Bibliography of the group for efficient algorithms at the University of Munich.
🗹 Mendeley	m	The digital library of Mendeley.
PubMed	L _M	A large and well-known resource for the life sciences including medicine and biology
🗹 Wiley	1	The digital library of the publisher Wiley.

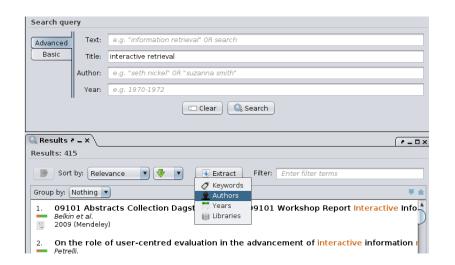
Starting: Search term completion



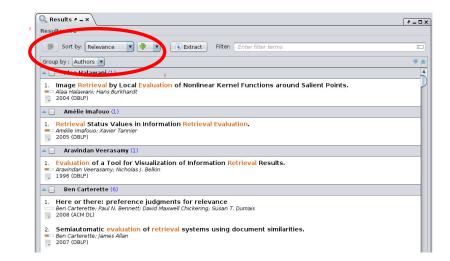
Starting: Related Terms

information retrieval		Search Options	•
language information retrieval	Ŷ	Explore concepts: informati	on retrieval +
information retrieval system		Natural Language Proces	conferences
language information retrieval based		queries	machine learning
enrich information retrieval		algorithm	search engines
modern information retrieval		Springer	theory

Starting: Identify important authors 'extract authors' in ezdl



Browsing: Sort/group results by different criteria



Browsing: Meaningful Surrogates

Any field contains I topology O Add sub-clause Add another rule Search	
Currently displaying 1 – 15 of 5882	Journals
The compact weak topology on a Banach space.	Commentationes Mathematicae 856 Universitatis Carolinae
Manuel González, Joaquín M. Gutiérrez — Extracta Mathematicae	Czechoslovak Mathematical 335 Journal
Throughout [this paper], E and F will denote Banach spaces. The bounded weak topology on	Annales de l'institut Fourier 310
a Banach space E, noted bw(E) or simply bw, is defined as the finest topology that agrees with the weak topology on bounded sets. It is proved in [3] that bw(E) is a locally convex topology if	Compositio Mathematica 282
and only if E is r	Algebraic \& Geometric Topology 276
A note on the topology associated with a locally convex space. Radenović, Stojan — Publications de I'institut Mathi'ematique. Nouvelle Si'erie	More ♥ Years
If R is a property in the class of locally convex spaces, which is invariant under passage to an	2011 2
arbitrary inductive limit and the finest locally convex topology, then for every locally convex	2010 93
space (E,t) there exists a locally convex topology Rt with the property R. For example, R is one	2009 103
of the pro	2008 124
Hausdorff topology and uniform convergence topology in spaces of	2007 135
continuous functions	More 🇭

Browsing: Highlighting in the Result List

Google	peer to peer information retrieval
Suche	Ungefähr 5.530.000 Ergebnisse (0,21 Sekunden)
Alles	Wissenschaftliche Artikel zu peer to peer information retrieval
Bilder	<u>Peer-to-peer information retrieval using self-organizing</u> - Tang - Zitiert durch: 541 mining for the biologist: from information retrieval to Jensen - Zitiert durch: 331 indexing for efficient peer-to-peer information retrieval - Tang - Zitiert durch: 179
Maps	indexing for enclent peer-to-peer information retreval - rang - Zittert durch. 179
Videos	[PDF] An Architecture for Peer-to-Peer Information Retrieval infoscience.epfl.ch//P2P-IR Architecture.pdf - Diese Seite übersetzen
News	Dateiformat: PDF/Adobe Acrobat - Schnellansicht von K Aberer - 2004 - Zitiert durch: 35 - Äbnliche Artikel
Shopping	An Architecture for Peer-to-Peer Information. Retrieval. Karl Aberer, Fabius Klemm,
Mehr	Martin Rajman, Jie Wu. School of Computer and Communication Sciences
	Workshop on Information Retrieval in Peer-to-Peer Networks (P2PIR
Dortmund Standort ändern	Isirwww.epfl.ch/p2pir2006/ - Diese Seite übersetzen Workshop on Information Retrieval in Peer-to-Peer Networks. collocated with the. ACM Fifteenth Conference on Information and Knowledge Management (CIKM

Chaining: Clickable Entries in Result Pages

Chaining: backward/forward chaining of references

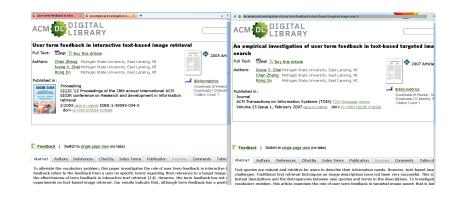


			\frown						
Abstract	Authors	References	Cited By	ndex Terms	Publication	Reviews	Comments	Table of Contents	
7 Citatio	ns								
Pro		of the 11th Pac						-oriented line drawi essing: Part I, Septe	ing image retrieval, ember 21–24, 2010,
								context for searchin 2, October 2010	ng Multimedia
								ased on weighted v 2, p.4301–4313, No	
								feedback in web inf er 18–20, 2006, Cop	
									model, Proceeding ish Columbia, Cana
		, Chen Zhang on Informatior					feedback in t	ext-based targeted i	mage search, ACM
		lli, On the role ent: an Intern					nteractive inf	ormation retrieval, I	Information Process

Forward Chaining in Web Searches

Google	link:www.is.inf.uni-due.de/courses/index.html
Suche	7 Ergebnisse (0,23 Sekunden)
Alles	DAFFODIL: Strategic Support Evaluated
Bilder	www.is.informatik.uni-duisburg.de/bib/docs/Klas_04ta.html DAFFODIL: Strategic Support Evaluated. UDEFakultäten
Maps	IngenieurwissenschaftenAbteilung InKolnformationssystemeDAFFODIL: Strategic Support Evaluated
Videos	Dil AQ: o Divitel Library Apposition Convision
News	DiLAS: a Digital Library Annotation Service www.is.informatik.uni-duisburg.de/bib/docs/Agosti_etal_05.html
Shopping	UDEFakultätenIngenieurwissenschaftenAbteilung InKolnformationssystemeDiLAS
Mehr	Teaching www.is.informatik.uni-duisburg.de//index.htm Diese Seite übersetzen
Das Web	UDEFakultätenIngenieurwissenschaftenAbteilung InKolnformationssysteme
Seiten auf Deutsch Übersetzte Seiten	Comparing different architectures for query routing in peer-to www.is.informatik.uni-duisburg.de/bib//Nottelmann_Fuhr_06.html Zitationsschlüssel: Nottelmann/Fuhr:06; Titel: Comparing different

Differentiating: comparison of result pages



Extracting: Highlighting in the result page

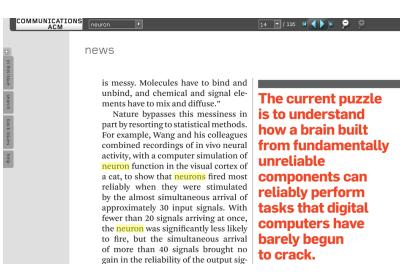
Extracting: Searching in the result page

Sheffield July 29

27th Annual International ACM SIGIR Conference Workshop on Peer-to-Peer Information Retrieval

SIGIR is the major international forum for the presentation of new research results and the demonstration of new systems and techniques in the broad field of information retrieval.

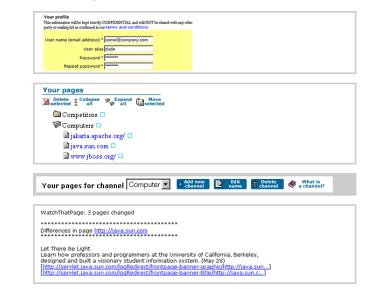
This SIGIR workshop on Peer-to-Peer Information Retrieval focus on new methods of resource representation, resource selection, and data fusion in peer-to-peer networks. The workshop particularly encourages papers that address heterogeneous peer-to-peer networks, as well as papers about methods that cope with partial and uncertain information. However, more broadly, papers are solicited on any topic related to information retrieval in peer-to-peer networks.



Monitoring 'Saved Searches' at PubMed

Table of Contents	My NCBI Home > Saved Data > Saved Searches
My NCBI Home	Saved Search Settings
My Saved Data	Your PubMed search
Search Filters	Search: phototherapy diabetic neuropathy
Preferences	Name of Search: phototherapy diabetic n
About My NCBI	E-mail: doctorpeabody@gmail.com
	Would you like e-mail updates of new search results?
	C No thanks. Yes, once a month. Which day? the first Saturdey * Yes, once a week. Which day? Monday * Yes, every day. Format: Report format: Summary * Number of items: Send at most: 5 items * C send even when there aren't any new results Any text you want to be added at the top of your e-mail (optional): Save

Monitoring Watchthatpage.com



Information management: Organizing and Annotating results

UNIVERSITĂT DUISBURG ESSEN			Aufstellungssystematik Zeitschriftenkatalog Sonderkataloge	Fernleihe Neuerwerbungen Feedback	<mark>Anmelden</mark> Neustart Hilfe	
----------------------------------	--	--	--	---	--	--

Dies ist eine temporäre Liste. Alle Titel und Listen werden am Ende der Sitzung gelöscht. Um Listen dauerhaft zu speichern, müssen Sie sich anmelden.

Merkliste

Merkliste (3)	•	Speichern/Senden	Lösch en	Listenverwaltung	

	#	\mathbf{X}	Urheber	Titel	Jahr	Bestand	Notiz
	1		Stock, Wolfgang G.	Information-Retrieval	2007	Campus Duisburg(4/4) Campus Essen(1/1)	Lehrbuch
e	2		Baeza-Yates, Ricardo	Modern information retrieval [Nachdr.]	2005	Campus Essen(4/2)	
•	3		Ferber, Reginald	Information Retrieval 1. Aufl.	2003	Campus Duisburg(1/1) Campus Essen(1/1)	

Universitätsbibliothek Duisburg-Essen Aleph-System © 2005 Ex Libris

Summary

Information management: personal library in Daffodil

8	😂 🎜 🛅 😠 🗃 🕄 🐯
0-	🛅 fuhr
0-	🛅 DELOS
o-	🛅 DPA
o-	🛅 FIZ Karlsruhe
o-	🛅 INEX
0-	🛅 Information Retrieval
0-	🛅 ir
۰-	NFS-EU
۰-	💼 patent
	E PGBetreuer
የ	The PHD
	🗢 🛅 Annotation-based Retrieval (Ingo)
	ዮ 🚋 Multithetic clustering (Gudrun)
	 Applications of matrix factorization for browsing/overview
	 Justification and interaction models
	የ 📷 Matrix factorization methods
	Socument clustering based on non-negative matrix factorization
	S Document clustering by concept factorization
	Multi-faceted Learning for Web Taxonomies
	📑 Multi-faceted Learning of Web Taxonomies
	Multi-faceted Learning of Web Taxonomies
	Multi-faceted Learning of Web Taxonomies http://en.wikipedia.org/wiki/Latent_semantic_analysis http://en.wikipedia.org/wiki/Probabilistic_latent_semantic_analysis
	Multi-faceted Learning of Web Taxonomies

Summary

- Interactive PRP for analysis and design of IIRS
- ► Information seeking behavior vs. searching
- cognitive models:
 - search as iterative process
 - large variation in search tasks
 - search influenced by many factors
- systems:
 - strategic support through high-level search functions (especially for typical cognitive actions)
 - proactive support
- user interface design on cognitive models

References I

Fuhr, N.

(2010).

Skriptum Information Retrieval, Kapitel 6 'Interaktives Retrieval'. http://www.is.inf.uni-due.de/courses/ir_ss10/folien/kapiir.pdf

Bates, M. J.

(1989).

The design of browsing and berrypicking techniques for the online search interface. Online Review 13(5), pp. 407–424.

Bates, M. J.

(1990).

Where Should the Person Stop and the Information Search Interface Start? *Information Processing and Management 26(5)*, pp. 575–591.

Belkin, N. J.

(1980).

Anomalous states of knowledge as a basis for information retrieval. *Canadian Journal of Information Science 5*, pp. 133–143.

References III

Hearst, M. A.

(2009). Search User Interfaces. Cambridge University Press.

Ingwersen, P.

(1992). Information Retrieval Interaction. Taylor Graham, London.

Ingwersen, P.

(1994).

Polyrepresentation of Information Needs and Semantic Entities, Elements of a Cognitive Theory for Information Retrieval Interaction. In: Croft, B. W.; van Rijsbergen, C. J. (Hrsg.): *Proceedings of the Seventeenth Annual International ACM SIGIR Conference on Research and Development in Information Retrieval*, pp. 101–111. Springer-Verlag, London, et al.

Kriewel, S.; Fuhr, N. (2007).

Adaptive Search Suggestions for Digital Libraries. In: Asian Digital Libraries: Looking Back 10 Years and Forging New Frontiers (ICADL 2007), pp. 220–229.

References II

Belkin, N.

(1996).

Intelligent information retrieval: Whose intelligence? In: *ISI '96: Proceedings of the Fifth International Symposium for Information Science*, pp. 25–31. Universitätsverlag Konstanz.

Cool, C.; Belkin, N. J.

(2002).

A Classification of Interactions with Information. In: Bruce, H.; Fidel, R.; Ingwersen, P.; Vakkari, P. (Hrsg.): *Emerging frameworks and methods. Proceedings of the Fourth International Conference on Conceptions of Library and Information Science (COLIS4)*, pp. 1–15. Libraries Unlimited, Greenwood Village.

Ellis, D. (1989).

A behavioural approach to information retrieval system design. Journal of Documentation 45(3), pp. 171–212.

Fuhr, N. (2008).

A Probability Ranking Principle for Interactive Information Retrieval. Information Retrieval 11(3), pp. 251–265. http://dx.doi.org/10.1007/s10791-008-9045-0.

References IV

Kriewel, S.; Fuhr, N.

(2010).

An evaluation of an adaptive search suggestion system.

In: 32nd European Conference on Information Retrieval Research (ECIR 2010).

Meho, L. I.; Tibbo, H. R. (2003).

Modeling the information-seeking behavior of social scientists: Ellis's study revisited. Journal of the American Society for Information Science and Technology 54(6).

pp. 570–587.

Robertson, S. E.

(1977).

The Probability Ranking Principle in IR. Journal of Documentation 33, pp. 294–304.

Schaefer, A.; Jordan, M.; Klas, C.-P.; Fuhr, N. (2005).

Active Support For Query Formulation in Virtual Digital Libraries: A case study with DAFFODIL.

In: Rauber, A.; Christodoulakis, C.; Tjoa, A. M. (Hrsg.): *Research and Advanced Technology for Digital Libraries. Proc. European Conference on Digital Libraries (ECDL 2005)*, Lecture Notes in Computer Science. Springer, Heidelberg et al.

References V



Using Eye-Tracking with Dynamic Areas of Interest for Analyzing Interactive Information Retrieval.

In: Proceedings of the 35th international ACM SIGIR conference on Research and development in Information Retrieval, pp. 1165-1166.

Turpin, A. H.; Hersh, W.

(2001). Why batch and user evaluations do not give the same results. In: Croft, W. B.; Harper, D.; Kraft, D. H.; Zobel, J. (Hrsg.): *Proceedings of the* 24th Annual International Conference on Research and development in Information Retrieval, pp. 225–231. ACM Press, New York.

Zuccon, G.; Azzopardi, L.; van Rijsbergen, C. J. (2011).

The interactive PRP for diversifying document rankings. In: Ma, W.-Y.; Nie, J.-Y.; Baeza-Yates, R. A.; Chua, T.-S.; Croft, W. B. (eds.): *SIGIR*, pages 1227–1228. ACM.