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Semantic Web Mining and the Representation, Analysis, and Evolution of Web Space

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More info: www.berendt.de

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The Web is mankind´s largest repository of knowledge ...

3

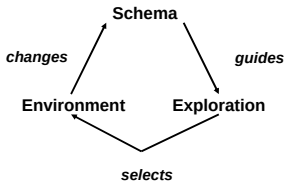
... but knowledge isn´t something that can be “put in a container and then used as the need arises“.

4

Knowledge is constructed in learning activities.

5

Knowledge$_1$ (“in people‘s minds“) is created by interaction with the environment (e.g., Neisser, 1967)



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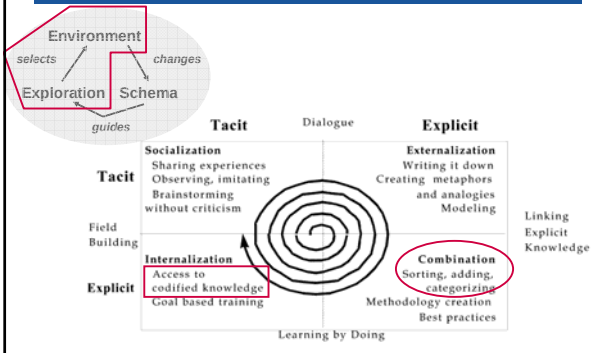
Knowledge$_2$ (codified) is part of the environment; Learning accesses this knowledge (e.g., Nonaka, 1991)



Semantic Web Mining

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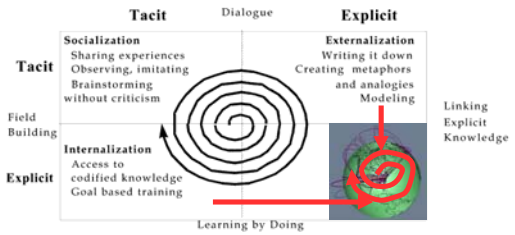

Approaches to the current Web's biggest challenges:
lots of data, human-understandable

8

Web Mining
extracts implicit knowledge

Semantic Web Mining
- **use semantics to improve mining**
- **use mining results to generate semantics**

The Semantic Web makes knowledge machine-understandable

[Berendt, Hotho, & Stumme, *Proc. ISWC* 2002]
[-"- (Eds.), *Proc. WS Semantic Web Mining at ECML/PKDD* 2001 and 2002]
[Berendt, Hotho, Mladenic, van Someren, Spiliopoulou, Stumme (Eds.), *Web Mining: From Web to Semantic Web*, 2004]

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Web Mining

(Semantic) Web

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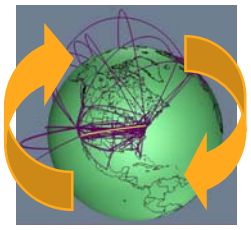
Web Mining

Semantic Web



Extracting semantics from Web content & structure – ideas and examples

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Using syntactic structure, semi-automatically learn

- **Ontologies (build or extend Yahoo-like taxonomies; Web-scale example *KnowItAll*: „... such as ...", see Etzioni et al. 2004)**
- **Instances of concepts and relations in a given ontology (ontology population)**
  - **Technique: Information extraction**
  - **From textual information including tables**
    - **Krátky, Andrt, & Svátek**
  - **From visual information including text layout**
    - **Burget; Gatterbauer, Krüpl, Holzinger, & Herzog; Hassan & Baumgartner; Labský, Vacura, & Praks**
  - **From structure (hyperlinks)**
    - **Frivolt & Bieliková**
- **Interactive learning**
  - **Ceresna; Schindler, Arya, Rath, & Slany**
- **Re-using existing conceptualizations**
  - **Švihla & Jelínek**

PS: This is just my understanding of your papers – please send me email if you find I've missed something!

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(Semantic) Web

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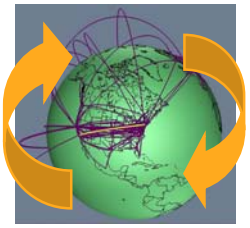
Web Mining

Semantic Web



Exploiting semantics for textual Web resources – ideas and examples

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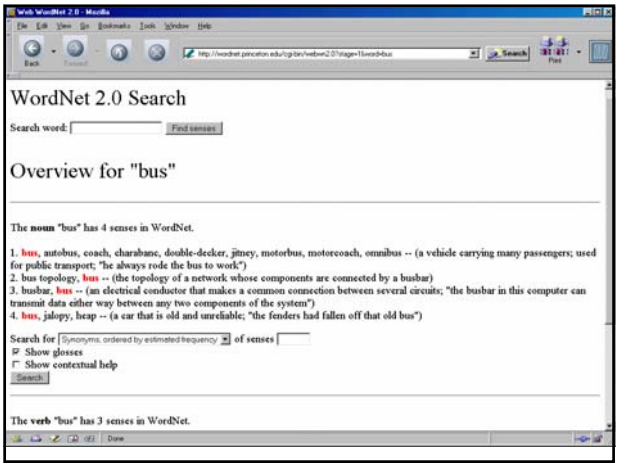
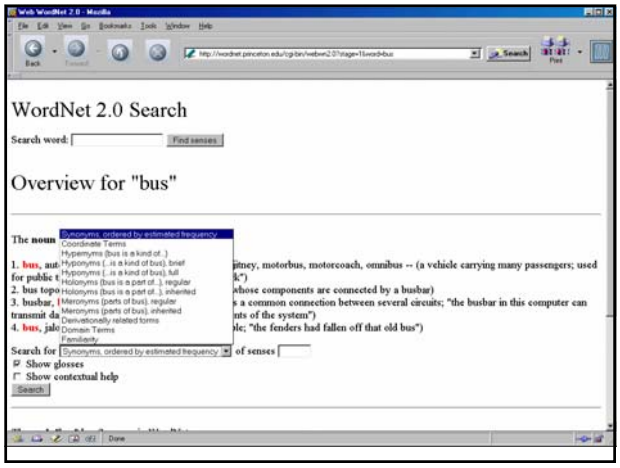
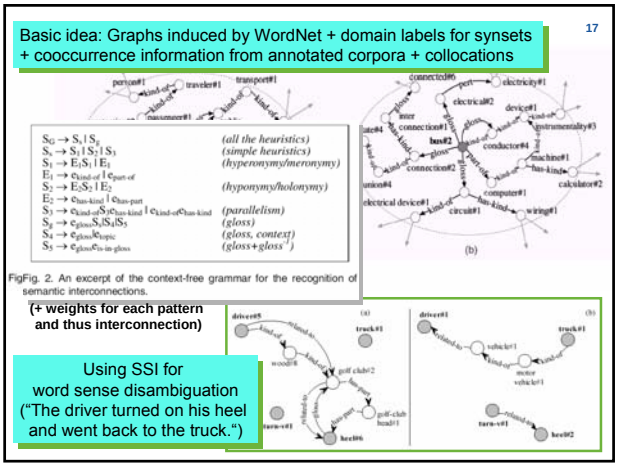
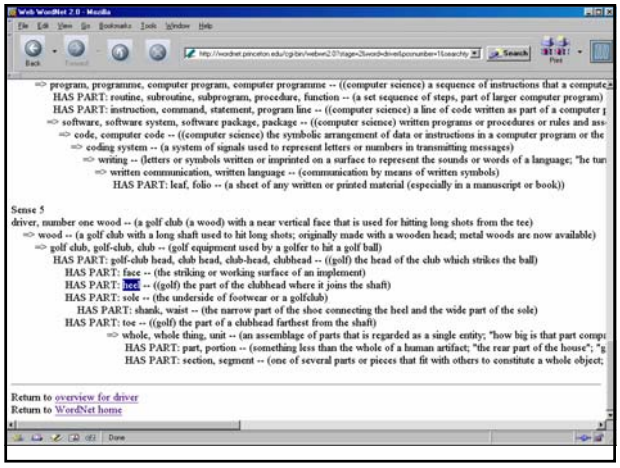
- Allowing expert users to contribute ontologies for semantics-enhanced IE
  - Schindler, Arya, Rath, & Slany
- Using ontologies to build templates for the composition of Web services
  - Svátek & Vacura
- Use ontologies as additional structure on the tokens in a text to
  - disambiguate meaning (e.g., word sense disambiguation: Navigli & Velardi, 2005)
  - reveal additional structure (e.g., clustering: Hotho, 2004)
  - help in the discovery of new ontological structures, or in instance learning (e.g., Navigli & Velardi, 2005; Hotho, 2004; KDD Cup 2005)

R. Navigli & P. Velardi. Structural Semantic Interconnections: a knowledge-based approach to word sense disambiguation. *IEEE Transactions on Pattern Analysis and Machine Intelligence* (27-7), 2005.









SSI for ontology learning

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1. Extract pertinent domain terminology:
   - Simple and multiword expressions that consistently occur in domain-related corpora and are not found in other domains (e.g., packet switching network)
2. Web search of available NL definitions from glossaries or documents
3. Use context-free grammar to
   1. filter out non-relevant definitions, based on statistical domain model
   2. parse definitions to extract kind-of information
4. Arrange terms in hierarchical trees
5. Link sub-hierarchies to the concepts of a core ontology (general-purpose: WordNet)
6. Provide the output to domain specialists for evaluation and refinement

```
S → PP ',' NP SEP
NP → N1 KIND1
KIND1 → MOD1 NOUN1
MOD1 → Verb | Adj | Verb ',' MOD1 | Adj ',' MOD1
NOUN1 → Noun
N1 → Art | Adj
SEP → ',' | '.' | Prep | Verb | Wh
PP → Prep NP
```

SSI: Word sense disambiguation and ontology learning

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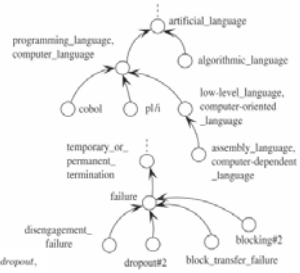
- ***artifical language*: monosemous in WordNet**
- ***temporary or permanent termination*: does not exist in WordNet**
- ***termination*: need to apply WSD (#1: end of a time span, #2: expiration of a contract)**
  - → **use terms that occur in the subtree and have a lexical correspondent in WordNet**
  - → $P = \{disengagement, failure, block, transfer, dropout, temporary, permanent\}$.



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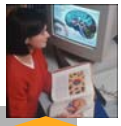
Web Mining

Semantic Web



Under-stand

```
...
p3ee24304.dip.t-dialin.net
[19/Mar/2002:12:03:51 +0100
/search.html?t=jane%20auste
3785&ord=asc HTTP/1.0" 200
p3ee24304.dip.t-dialin.net
[19/Mar/2002:12:05:06 +0100
/search.html?t=jane%20austen&m=vide
o&SID=023785&ord=desc HTTP/1.0" 200
8450
p3ee24304.dip.t-dialin.net - -
[19/Mar/2002:12:06:41 +0100] "GET
/view.asp?id=3456&SID=023785
HTTP/1.0" 200 3478
...
```

Application: Search in knowledge portals

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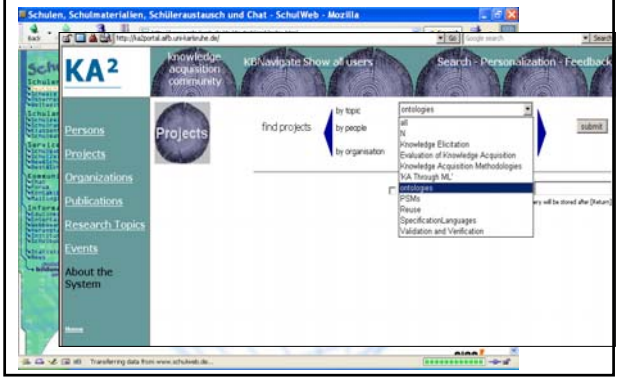

Ontology-based modelling of behaviour: URLs and application events

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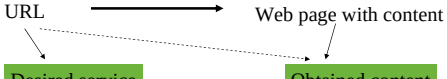

Definition 24.3.1 *A* site model $M := (\mathcal{S}, \mathcal{C}[, \mathcal{D}_S][, \mathcal{D}_C])$ *is a tuple consisting of a service ontology* $\mathcal{S}$*, a content ontology* $\mathcal{C}$*, and optionally tuples* $\mathcal{D}_S$ *and/or* $\mathcal{D}_C$ *consisting of the service/content information dimensions.*

Definition 24.3.2 *An* atomic application event *for a site model* $M = (\mathcal{S}, \mathcal{C}[, \mathcal{D}_S][, \mathcal{D}_C])$ *is a tuple* $AAE := (S, C)$ *consisting of (1)* S*, the service: a concept or relation from the service ontology* $\mathcal{S}$*, or (if and only if* $\mathcal{D}_S$ *is present in* M*) a tuple whose components are concepts/relations from* $\mathcal{S}$ *describing the service along the dimensions of* $\mathcal{D}_S$*, and (2)* C*, the content: a concept or relation from the content ontology* $\mathcal{C}$*, or (if and only if* $\mathcal{D}_C$ *is present in* M*) a tuple whose components are concepts/relations from* $\mathcal{C}$ *describing the service along the dimensions of* $\mathcal{D}_C$*.*

Definition 24.3.3 *A complex application event CAE is a nonempty (1) set, (2) sequence (i.e., order relation), or (3) another sequential structure (a regular expression, a context-free grammar, ...), whose elements are atomic application events.*

Berendt, B., Stumme, G., & Hotho, A. (2004). Usage mining for and on the Semantic Web. In H. Kargupta, A. Joshi, K. Sivakumar, & Y. Yesha (Eds.), *Data Mining: Next Generation Challenges and Future Directions*. Menlo Park, CA: AAAI/MIT Press.

Semantics of requests
Step 1: Domain ontology

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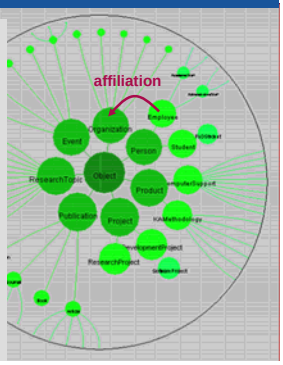
- community portal ka2portal.aifb.uni-karlsruhe.de
- ontology-based:
  - Knowledge base in F-Logic
  - Static pages: annotations
  - Dynamic pages: generated from queries & KB
  - Queries also in F-Logic
  - Logs contain these queries



[Oberle, Berendt, Hotho, & Gonzalez, *Proc. AWIC* 2003]

Semantics of requests
Step 2: Modelling requests as atomic application events

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RESEARCHER
PERSON
PROJECT
PUBLICATION
RESEARCHTOPIC
EVENT
ORGANIZATION
RESEARCHINTEREST
LASTNAME
TITLE
ISABOUT
EVENTS
EVENTTITLE
WORKSATPROJECT
AUTHOR
AFFILIATION
ISWORKEDONBY
PROGRAMCOMMITTEE
EMPLOYS
NAME
RESEARCHGROUPS
EMAIL

An example query with concepts and relations:

FORALL N,PEOPLE <-PEOPLE:
Employee[affiliation->> "http://www.anInstitute.org"]
and PEOPLE:Person[lastName->>N].

Query =
feature vector of concepts + relations

➔

Clustering, Association rules, Classification, ...

Session =
feature vector of concepts + relations, summed over all queries in the session

Semantics of sequences
Step 3: Using ontologies of behaviour for info. Extraction – Modelling sequences as composite application events

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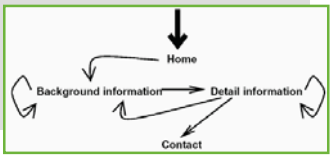
Composite application events - Example *customer typology*

- Based on background theory from marketing: the customer buying cycle
- Modelled in terms of regular expressions and employed in Web usage mining
- Example: *knowledge builders*
  (as opposed to, e.g., *direct buyers*)



[Moe, *Journal of Consumer Psychology*, 2002]
[Spiliopoulou, Pohle, and Teltzrow, *Proc. Wirtschaftsinformatik 2002*]

2. Semantics of sequences
for Step 3: an interactive tool with a query language

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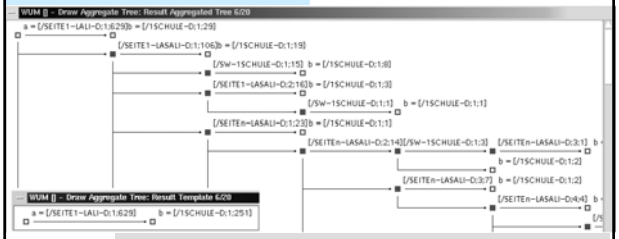
```
select t
from node a b, template a * b as t
where a.url startswith "SEITE1-"
and a.occurrence = 1
and b.url contains "1SCHULE"
and b.occurrence = 1
and (b.support / a.support) >= 0.2
```



Tool: www.hypknowsys.de; Data: [Berendt & Spiliopoulou, *VLDB Journal*, 2000]

Semantics of sequences
Step 4: Pattern discovery / instance learning

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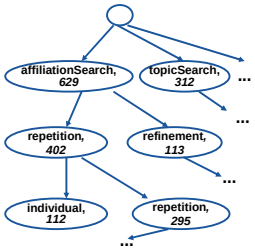
An ontology of composite application events (CAEs)

- Define templates as regular expressions
  - of atomic application events
  - of transitions (between atomic application events)

  Ex. [.search .* individual]
- Discover instances by learning a CAE trie



[Berendt & Spiliopoulou, *VLDB Journal*, 2000]
[Berendt, *Data Mining and Knowledge Discovery*, 2002]

Semantics of sequences
Step 5: Pattern evaluation

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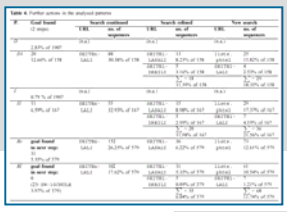
Use pattern statistics to

- derive descriptive measures of CAEs
  - support, confidence
  - popularity, effectiveness, efficiency
- apply inferential statistics to compare CAEs



[Berendt, *Data Mining and Knowledge Discovery*, 2002]

Communication – Visual data mining

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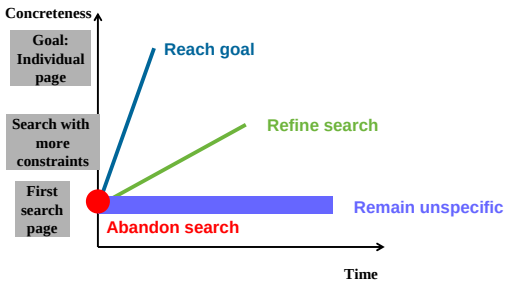
Step 6: Mapping an ontological relation over concepts to a linear order; mapping to visual variables



Communication – Visual data mining

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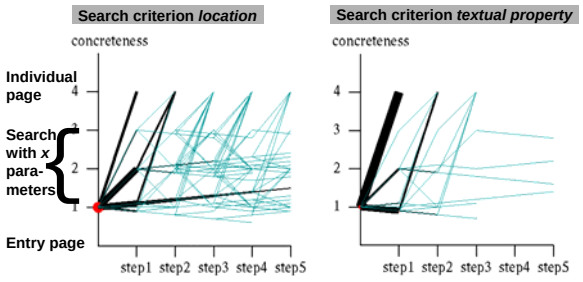
Step 6 – Example



[Berendt, *Data Mining and Knowledge Discovery*, 2002], [Berendt, *Postproc. WebKDD 2001*]

Communication – Visual data mining

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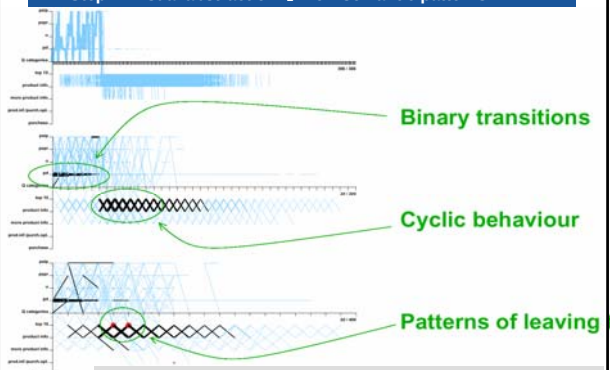
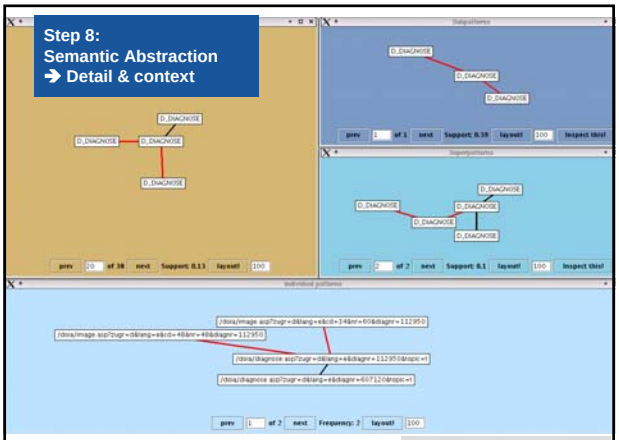
Step 7: Visual abstraction ➔ new semantic patterns



Data:see [Berendt, Günther, & Spiekermann, *Communications of the ACM*,2005]

Step 8:
Semantic Abstraction
➔ Detail & context



[Berendt, *Proc. WebKDD 2005*]

Case study: Information search in a medical portal*

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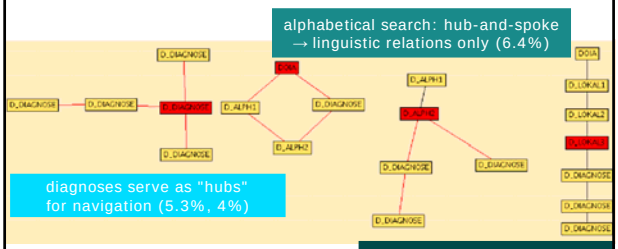

*** (20333 requests / 1397 sessions from Web log collected in 2001/2002; preprocessing: mining in concept space, see paper in proceedings)**

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Web Mining

(Semantic) Web

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Web Mining

Semantic Web

contribute

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Application: Knowledge construction for educational portals / Digital Libraries

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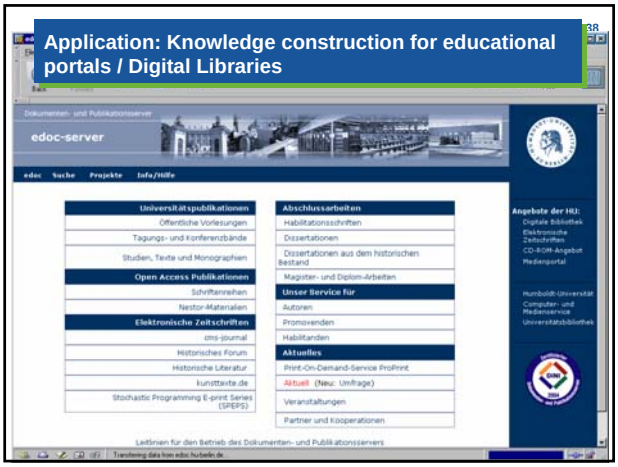

Knowledge contributions: Data and metadata

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Dissertation Markup Language DiML

40

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Authoring support for document servers

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- **Surveys (ca. 2500 persons; 12-14% response rate) & Web usage mining (ca. 11000 sessions) showed:**
  - **Metadata creation is one of the main barriers for contribution.**
- **Reasons include deficiencies in**
  - **information flow** → **Marketing**
  - **understanding and use of structured search**
  - **education in structured writing** → **Education**
  - **HCI aspects**

Intelligent Authoring Tools

[Berendt, Brenstein, Li, & Wendland, *Proc. ETD* 2003]
[Berendt, *Proc. AAAI Spring Symposium KCVC*, 2005]

Consequences of metadata neglect

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Why is this a problem?

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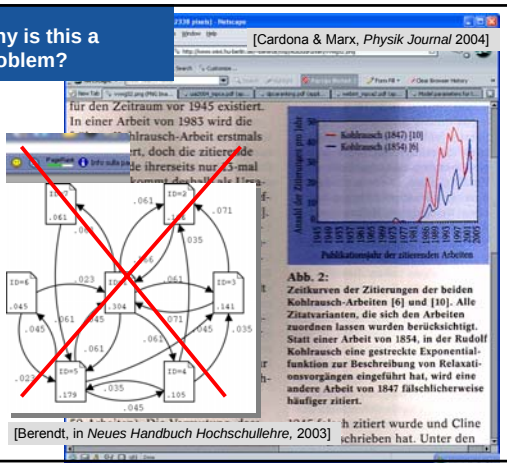

System architecture

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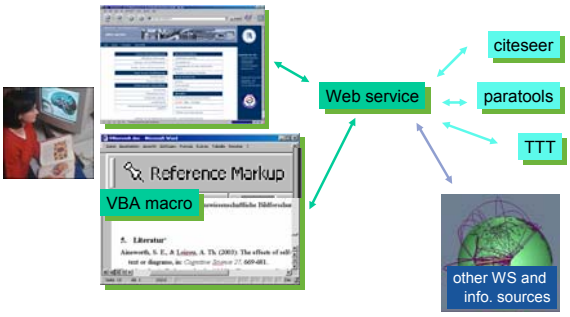

Usage interface

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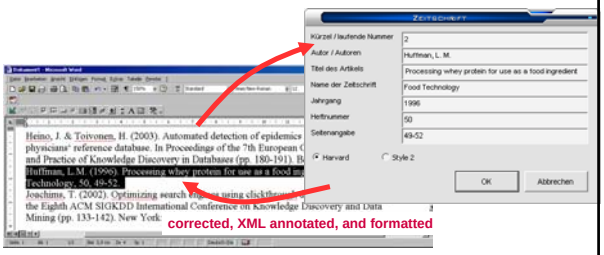
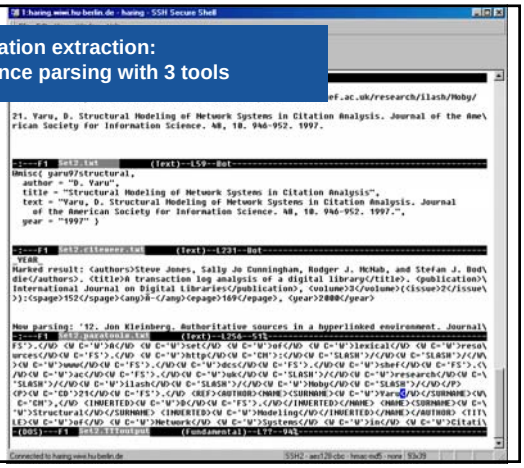


Information extraction: Reference parsing with 3 tools



Paratools-Zitations-Parsing

http://paracite.eprints.org

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A database of templates of the form

'_AUTHORS_ (_YEAR_). _TITLE_. _PUBLICATION_, _VOLUME_(_ISSUE_):_PAGES_'

each _XXX_ is associated with a regular expression

- Ex.: _YEAR_ → ([[:digit:]]{4})

2 weighting factors

- reliability: how "syntactically fixed" is a regular expression?
  - Ex.: _URL_ > _TITLE_
- concreteness = number of fixed symbols
  - Ex.: '_AUTHORS_, _PUBLICATION_, in press' > '_AUTHORS_, _PUBLICATION_'

Templates are matched against the reference.

Choose the templiate with the highest *reliability*, or (if these are equal) with the highest *concreteness*.

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Outlook 1: Diversity
(or: Web space and real-life spaces)

Which diagnosis is that?

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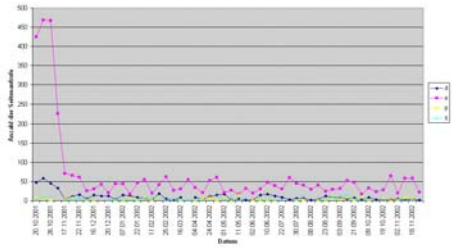
Request frequency for a specific diagnosis in the investigated eHealth portal, depending on time and request language



[Yihune, 2003]

Hypotheses: search preferences

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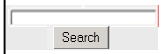
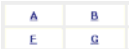
Search option	Characteristics	Presumably preferred by
Search engine 	■ little context ■ fast information access ■ no hierarchies	■ Low context ■ Low Uncertainty Avoidance ■ Short-Term oriented ■ Low Power Distance
Alphabetically organized links 	■ large hierarchies	■ High Power Distance
Content-organized links 	■ highest amount of (context) information ■ more time-consuming information access ■ large hierarchies	■ High context ■ High Uncertainty Avoidance ■ Long-Term oriented ■ High Power Distance

(Kralisch & Berendt, *Proc. IWIPS 2004)*

Search behaviour: sample results

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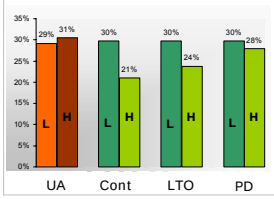
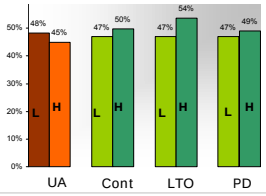
1. **Which search options were used?**
   - Expected results
   - Unexpected results
   - all results significant ($p<0.001$)

UA – Uncertainty Avoidance
Cont – Context Specifity
LTO – Long-Term Orientation
PD – Power Distance

search engine:



content-organized links:



Interactions between language and domain knowledge

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expected:

Search option	Characteristics: language use	Characteristics: required terminological knowledge	L1 users	L2 users
Search engine	productive	extensive		Language-sensitive search behaviour
Alphabetical search			H3	
Content-organized search	receptive	limited		H2

observed:

Search option	L1 users with High DK	L1 users with Low DK	L2 users with High DK	L2 users with Low DK
Search engine				
Alphabetical search				Language-sensitive search behaviour
Content-organized search				

Kralisch & Berendt, *New Review of Hypermedia and Multimedia*, in press)

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Outlook 2: Community

bibster.semanticweb.org

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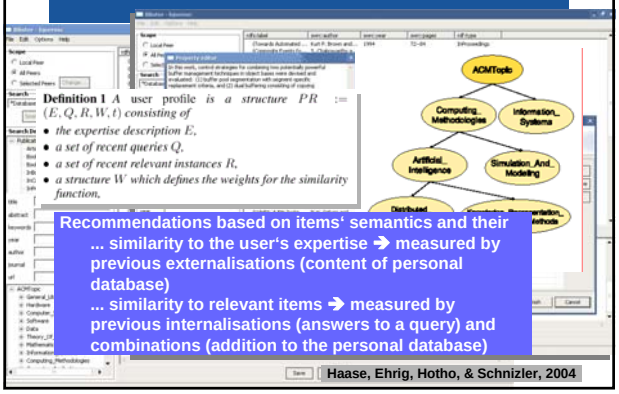
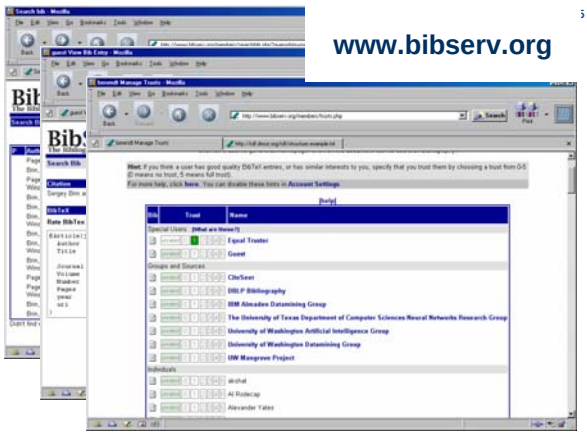


Recommendations based on items' semantics and their
... similarity to the user's expertise ➔ measured by previous externalisations (content of personal database)
... similarity to relevant items ➔ measured by previous internalisations (answers to a query) and combinations (addition to the personal database)

Haase, Ehrig, Hotho, & Schnizler, 2004



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Outlook 3: Fun!

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?
Outlook 4: Share the initiative – automated Web service search and composition for SWM
?

Thank you for your attention!