

## 8 Ranked Retrieval of XML Data

- 8.1 Basics of XML and XPath
- 8.2 Search with Ontological Similarities (XXL, COMPASS)
- 8.3 Search with Structural Similarities (XSearch)
- 8.4 Text Adjacency Search (XRank)

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.1

## 8.1 Basic Concepts of XML

- (Freely definable) tags: book, title, author
  - with start tag: <book> etc.
  - and end tag: </book> etc.
- **Elements:** <book> ... </book>
  - Elements have a name (book) and a content (...)
  - Elements may be nested.
- Each XML document has a root element and forms a tree.

Element content may be typed (mostly PCDATA – parsed character data, i.e., strings, possibly with nested elements). Elements may have **attributes** that have a name and a value (content), e.g. <article year=1999>.

Elements optionally have id attributes (element ids) from which references within a document can be constructed via idref attributes.

Elements may have outgoing hyperlinks via href attributes.

Elements with a common parent are ordered.

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.2

## XML: Fancy Example (from T. Grust)



```
<strip copyright=„...“ year=2000>
<panel no=„1“>
  <speech speaker=„Pointy-Haired Boss“ to=„Dilbert“>
    <character> Pointy-Haired Boss </character>
    <bubble> I think we should build an SQL database. </bubble>
  </speech> </panel>
<panel no=„2“>
  <speech mode=„thinking“>
    <character> Dilbert </character>
    <bubble> Does he understand what he said or ... </bubble>
  </speech> </panel>
<panel no=„3“>
  <speech speaker=„Dilbert“ mode=„question“>
    <bubble> What color do you want that database? </bubble>
  </speech>
  <speech speaker=„Pointy-Haired Boss“ tone=„self-confident“>
    <bubble> I think mauve has the most RAM.</bubble>
  </speech> </panel>
</strip>
```

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.3

## IMDB Data in XML

```
<?xml version="1.0" encoding="ISO-8859-1" ?>
<movie id="62480">
  <title>Contact</title>
  <production_year>1997</production_year>
  <production_country>USA</production_country>
  <production_language>English</production_language>
  <production_language>Spanish</production_language>
  <production_location>Arecibo Observatory, Arecibo, Puerto Rico</production_location>
  <production_location>Canyon de Chelly National Monument, Chinle, Arizona, USA</prod
...
<producer xmlns:xlink="http://www.w3.org/1999/xlink/namespace" xlink:type="simple"
  <name>Bradshaw, Joan</name> <job>executive producer</job>
</producer>
<cast order="credits">
<casting position="1">
<actor xmlns:xlink="http://www.w3.org/1999/xlink/namespace" xlink:type="simple" xlin
  Foster, Jodie</actor>
<role>Dr. Eleanor Ann 'Ellie' Arroway</role>
</casting>
...
<plot author="Ed Howell <on@combase.com">Contact, based on the novel of the
  same name by Carl Sagan, is the story of a free thinking radio astronomer ... </plot>
...
<goof type="factual error">: It is impossible for Dr. Arroway to have graduated from
  MIT Magna Cum Laude, as MIT gives no such distinctions.</goof>
```

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.4

## XML-based Data Annotation Standards

XML is the basis for a variety of application-field-specific data annotation standards, e.g.:

- MathML: Mathematical Markup Language
- CML: Chemical Markup Language
- CellML for computer-based biological models
- ebXML for e-business data (e.g., invoices)
- WeatherML for weather observation data
- NITF for news agencies
- etc. etc.

XML is mere syntax, but it creates a momentum towards standardized terminologies (ontologies), thus potentially enabling large-scale data exchange (and more effective information search)

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.5

## CML Example

```
<cml title=„ethanol“ id=„cml_ethanol_karne“>
<molecul title=„ethanol“ id=mol_ethanol_karne“>
  <formula> C2 H6 O </formula>
  <string title=„CAS“>64-17-5</string>
  <float title=„molecular weight“>46.07</float>
  <atomArray>
    <atom id=„ethanol_karne_a_1“>
      <float builtin=„x3“ units=„A“>1.0303</float>
      <float builtin=„y3“ units=„A“>0.8847</float>
      <float builtin=„z3“ units=„A“>0.9763</float>
      <string builtin=„elementType“>C</string>
    </atom>
    <atom id=„ethanol_karne_a_2“>
      ... </atom> ...
  </atomArray>
  <bondArray>
    <bond id=„ethanol_karne_b_1“>
      <string builtin=„atomRef“>ethanol_karne_a_1</string>
      <string builtin=„atomRef“>ethanol_karne_a_2</string>
      <string builtin=„order“ convention=„MDL“>1</string>
    </bond>
    ...
  </bondArray>
```

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.6

## Boolean Retrieval with XPath and XQuery

XPath and XQuery are query languages for XML data, both standardized by the W3C and supported by various database products. Their search capabilities include

- **logical conditions** over element and attribute content (first-order predicate logic a la SQL; simple conditions only in XPath)
  - **regular expressions** for pattern matching of element names along paths or subtrees within XML data
- + joins, grouping, aggregation, transformation, etc. (XQuery only)

In contrast to database query languages like SQL an XML query does not necessarily (need to) know a fixed structural schema for the underlying data.

A **query result** is a set of qualifying nodes, paths, subtrees, or subgraphs from the underlying data graph, or a set of XML documents constructed from this raw result.

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.7

## XPath by Examples

<code>/movie/casting/actor</code>	all actors in the castings of all movies
<code>/movie[title=„Contact“]/casting</code>	all people in the casting of a given movie
<code>/movie[title=„Contact“]/actor</code>	all actors in a given movie
<code>/movie[title=„Contact“]/* /actor</code>	
<code>/movie/casting[@position=1]/actor</code>	all stars
<code>/movie[casting/actor = „Foster, Jodie“]/title</code>	titles of movies with a given actor
<code>/movie[casting/actor = „Foster, Jodie“]/casting[@position=1]/actor</code>	stars of movies with a given actor
<code>/movie[//goof]/casting/*</code>	casting details for movies with goofs
<code>/movie/casting/*[ancestor::*[name() = goof]]</code>	

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.8

## Semantics of XPath Queries

An XPath **path expression** (the core of a query) is a sequence of **location steps**, separated by /, each of which has

- a navigation axis (e.g., children denoted by /, descendants //, etc.) relative to a context node (i.e., a current node) and a
- condition to be matched, which may in turn be a path expression with a logical condition for the end node of a qualifying path

The evaluation of a path expression computes a function  $\text{nodes} \rightarrow 2^{\text{nodes}}$ ,

i.e., determines for a given initial context node the set of nodes that are reachable by the given sequence of location steps and whose paths satisfy all specified conditions.

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.9

## XPath Location Axes

The general form of a location step is

**axis::test[predicate]**

where test is a function on a node (with Boolean result, e.g., referring to the element name or position) and predicate is a function or a path expression

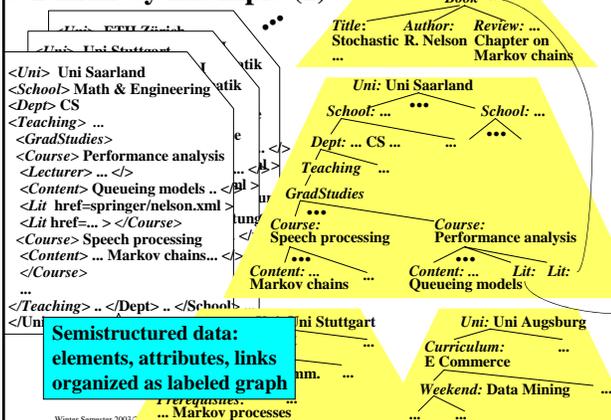
Axis	Shortcut	Comment
child::node()		node() is true for all nodes
descendant::node()	//	
descendant-or-self::node()		
self::node()	.	
parent::node()	..	
ancestor::node()		
ancestor-or-self::node()		
following::node()		in preorder traversal of doc tree
preceding::node()		
following-sibling::node()		siblings to the right
preceding-sibling::node()		siblings to the left
attribute::	@	

Winter Semester 2003/2004

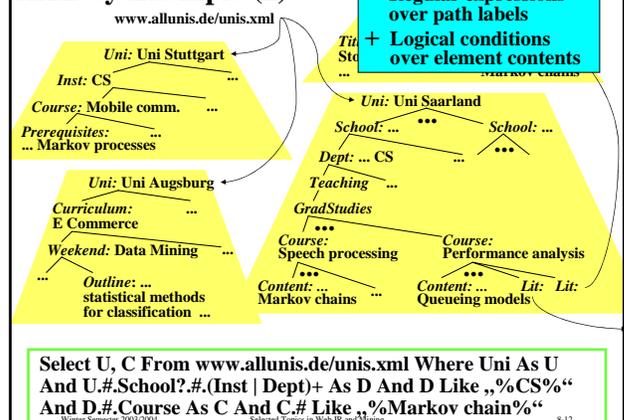
Selected Topics in Web IR and Mining

8.10

## 8.2 XXL by Example (1)



## XXL by Example (2)



### XXL by Example (3)

www.allunis.de/unis.xml

**Uni: Uni Stuttgart**  
Inst: CS ...  
Course: Mobile comm. ...  
Prerequisites: ... Markov processes

**Uni: Uni Saarland**  
School: ...  
Dept: ... CS  
Teaching ...  
GradStudies ...  
Course: Speech processing  
Content: ... Markov chains  
Course: Performance analysis  
Content: ... Lit: Lit: Queuing models

**Uni: Uni Augsburg**  
Curriculum: E Commerce  
Weekend: Data Mining ...  
Outline: ... statistical methods for classification ...

**Book**  
Title: Stochastic R. Nelson  
Author: Chapter on Markov chains  
Review: ...

Select U, C From www.allunis.de/unis.xml Where Uni As U And U.#.School?.#.(Inst | Dept)+ As D And D Like „%CS%“ And D.#.Course As C And C.# Like „%Markov chain%“

Winter Semester 2003/2004 Selected Topics in Web IR and Mining 8.13

### XXL by Example (4)

www.allunis.de/unis.xml

**Uni: Uni Stuttgart**  
Inst: CS ...  
Course: Mobile comm. ...  
Prerequisites: ... Markov processes

**Uni: Uni Saarland**  
School: ...  
Dept: ... CS  
Teaching ...  
GradStudies ...  
Course: Speech processing  
Content: ... Markov chains  
Course: Performance analysis  
Content: ... Lit: Lit: Queuing models

**Uni: Uni Augsburg**  
Curriculum: E Commerce  
Weekend: Data Mining ...  
Outline: ... statistical methods for classification ...

**Book**  
Title: Stochastic R. Nelson  
Author: Chapter on Markov chains  
Review: ...

Select U, C From www.allunis.de/unis.xml Where Uni As U And U.# As D And D ~ „CS“ And D.#.~Course As C And C.# ~ „Markov chain“

Winter Semester 2003/2004 Selected Topics in Web IR and Mining 8.14

### XXL by Example (5)

www.allunis.de/unis.xml

**Uni: Uni Stuttgart**  
Inst: CS ...  
Course: Mobile comm. ...  
Prerequisites: ... Markov processes

**Uni: Uni Saarland**  
School: ...  
Dept: ... CS  
Teaching ...  
GradStudies ...  
Course: Speech processing  
Content: ... Markov chains  
Course: Performance analysis  
Content: ... Lit: Lit: Queuing models

**Uni: Uni Augsburg**  
Curriculum: E Commerce  
Weekend: Data Mining ...  
Outline: ... statistical methods for classification ...

**Result ranking of XML data based on semantic similarity**

Select U, C From www.allunis.de/unis.xml Where Uni As U And U.# As D And D ~ „CS“ And D.#.~Course As C And C.# ~ „Markov chain“

Winter Semester 2003/2004 Selected Topics in Web IR and Mining 8.15

### XXL Concepts

Extensible, simple core language

Where clause: conjunction of regular with binding of variable

Elementary conditions on element/attribute names and contents

Select D, L, S From www.allunis.de/unis.xml Where Uni.#.School?.#.(Inst|Dept) As D And D.#.Lecturer As L And D.#.Student As S And L.Name = S.Name And L.Area Like „%XML%“

Semantic similarity conditions on names and contents

... D.#.~Lecturer As L And L.~Area ~ „XML“

Based on tf\*idf similarity of contents, ontological similarity of names, probabilistic combination of conditions

**Query Semantics:**  
• query is a path/tree/graph pattern  
• results are isomorphic paths/subtrees/subgraphs of the data graph

Winter Semester 2003/2004 Selected Topics in Web IR and Mining 8.16

### XXL Result Ranking

**Query:**  
Where Uni.#.School?.#.(Inst | Dept) As D And I.#.~Lecturer As L And L.~Area ~ „XML“

**Query Semantics:**  
• query is a pattern with relaxable conditions  
• results are approximate matches to query with similarity scores

**Data graph:**

```

graph TD
    Uni[Uni: UniDo] --> Dep[Dep: Inf]
    Uni --> Dep[Dep: Math]
    Uni --> Prof[Prof: N. Fuhr]
    Prof --> Teaching[Teaching]
    Prof --> Project[Project: IR on semistruct. data]
    Prof --> Project[Project: digital libr.]
    Teaching --> Lect[Lect: IR]
    Teaching --> Seminar[Seminar: XML]
  
```

**Result graph:**

```

graph TD
    Uni[1.0 Uni: UniDo] --> Dep[1.0 Dep: Inf]
    Uni --> Prof[0.9 Prof: N. Fuhr]
    Prof --> Project[0.8 Project: IR on semistruct. data]
    Prof --> Project[0.6 Project: digital libr.]
  
```

Relevance score: 0.432

Winter Semester 2003/2004 Selected Topics in Web IR and Mining 8.17

### XXL Semantics (1): Exact Queries and Variable-Free Path Expressions

Consider a data graph  $G=(V,E)$  with XML elements as nodes  $V$  and parent-child relationships or links as edges  $E$ .

An *XXL query* (actually its *Where clause*) is a conjunction of

- *search conditions*  $q_1, \dots, q_m$  ( $m \geq 1$ ) where each condition is a *path expression*, which can optionally be bound to an *element variable*, and
- *elementary content conditions*  $c_1, \dots, c_k$  ( $k \geq 0$ ) of the form *variable op constant* or *variable op variable* with comparison operator  $op \in \{=, !=, <, >, \text{Like}, \dots\}$ .

A *path expression* is a restricted regular expression over element labels: every label  $x$  is an expression, the wild card  $\#$  is an expression, and for expressions  $x$  and  $y$ ,  $x.y$ ,  $x|y$ ,  $x.\#y$  are expressions, too.

A path  $e_1 \dots e_n$  of the data graph satisfies a variable-free path expression  $q$  if the sequence of labels along the path is in the language defined by the regular expression  $q$ .  $\mu[q] \subseteq V^+$  denotes the *set of matching paths*.

Winter Semester 2003/2004 Selected Topics in Web IR and Mining 8.18

## XXL Semantics (2): Exact Queries with Variables

Every path expression can have an optional *As* clause which binds the end points of the qualifying paths to an element variable.

The „uses“ relation between path expressions  $q_1, \dots, q_m$  of the same query is defined as follows:  $q_i < q_j$  ( $q_j$  uses  $q_i$ ) if  $q_j$  contains a variable that is bound to the qualifying paths of  $q_i$ . We restrict the uses relation such that its transitive closure is irreflexive and acyclic.

A path expression  $q_i$  may contain element variables. The elements that are bound to the variables are substituted into  $p$ .

For a given variable binding  $v: \text{VAR} \rightarrow V$ , the result  $\mu_v[q_i] \subseteq V^+$  of  $q_i$  containing variables  $x, y, \dots$  is  $\mu[q_i[x/v(x).name, y/v(y).name, \dots]]$ .

A subgraph of  $G$  is a **result of the query** with path expressions  $q_1, \dots, q_m$  and elementary content conditions  $c_1, \dots, c_k$  with variables  $x, y, \dots$

if there is a (global) variable binding  $v$  such that the subgraph is the union of paths  $p_1, \dots, p_m$  such that

- $p_i$  is an approximate result in  $\mu_v[q_i]$  for all  $i$  and
- $c_j[x/v(x).content, y/v(y).content, \dots]$  evaluates to true for all  $j$ .

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8-19

## XXL Semantics (3): Queries with Similarity Conditions

Assume that similarity functions are defined between element names and between texts (and between dates, spatial names, etc.)

An element with label  $l$  **approximately matches** subcondition  $\sim\text{label}$  in path expression  $q_i$  if the similarity  $\text{sim}(l, \text{label}) > 0$ .

An element with content  $c$  bound to variable  $x$  approximately matches elementary content condition  $x \sim \text{const}$  if  $\text{sim}(c, \text{const}) > 0$ , and two elements with contents  $c, c'$  bound to variables  $x, x'$  approximately satisfy elementary content condition  $x \sim x'$  if  $\text{sim}(c, c') > 0$ .

A subgraph of  $G$  is an **approximate result** of query  $q$  with path expressions  $q_1, \dots, q_m$  and elementary content conditions  $c_1, \dots, c_k$  with variables  $x, y, \dots$

if there is a (global) variable binding  $v$  such that the subgraph is the union of paths  $p_1, \dots, p_m$  such that

- $p_i$  is an approximate result in  $\mu_v[q_i]$  for all  $i$  and
- $c_j[x/v(x).content, y/v(y).content, \dots]$  is approximately satisfied for all  $j$ .

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8-20

## XXL Semantics (4): Query Result Scoring

An **element  $e$  or a path  $p$  is scored**

- with regard to a subcondition of the form  $x, \#, x/y, \sim x$  by the similarity with which it approximately matches the subcondition, and an **element  $e$  or a pair  $(e, e')$  of elements is scored**
- with regard to an elementary content condition  $x \sim \text{const}$  or  $x \sim x'$  by the similarity between  $e$  and the given constant or between  $e$  and  $e'$ .

An approximately matching **subgraph is scored** with regard to a query by the product

of the scores of its components with regard to the underlying path subconditions and elementary content conditions.

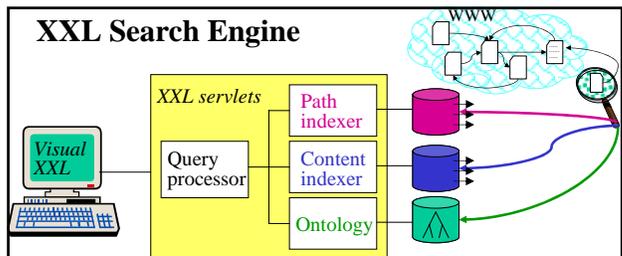
The **result of an XXL query with similarity conditions** is a ranked list of approximately matching subgraphs in descending order of scores.

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8-21

## XXL Search Engine



Select ... Where  
 Uni.#.(Inst|Dept) As F  
 And F ~ „Computer Science“  
 And F.#.~Course.#  
 ~ „Markov Chains“

- Query decomposition into index-supported subexpressions
- wide range of optimizations

Uni.#.(Inst|Dept) As F  
 F ~ „Computer Science“  
 F.#.Course.#  
 ~ „Markov Chains“  
 F.#.Seminar.#  
 ~ „Markov Chains“

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

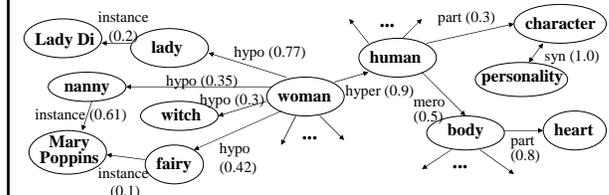
8-22

## Example Ontology

**woman, adult female – (an adult female person)**  
 => amazon, virago – (a large strong and aggressive woman)  
 => donna -- (an Italian woman of rank)  
 => geisha, geisha girl -- (...)  
 => lady (a polite name for any woman)  
 ...  
 => wife – (a married woman, a man’s partner in marriage)  
 => witch – (a being, usually female, imagined to have special powers derived from the devil)

## Ontology Graph

An ontology graph is a directed graph with concepts (and their descriptions) as nodes and semantic relationships as edges (e.g., hypernyms).



$$\text{sim}(c_1, c_2) = \frac{2| \{ docs \text{ with } c_1 \} \cap \{ docs \text{ with } c_2 \} |}{| \{ docs \text{ with } c_1 \} | + | \{ docs \text{ with } c_2 \} |}$$

**Dice coefficient from large corpus (e.g. focused crawl)**

$$\text{sim}^*(c_1, c_n) = \max_{i=1..n-1} \{ \prod_{i=1..n-1} \text{sim}(c_i, c_{i+1}) \mid \text{all paths from } c_1 \text{ to } c_n \}$$

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8-24

## Benefit from Ontology Service

Ontology service accessible via SOAP or RMI  
 Ontology filled with WordNet, geo gazetteer,  
 focused crawl results, extracted tables & forms

### Threshold-based query expansion:

substitute ~w by  $\{c_1 | \dots | c_k\}$  with all  $c_i$  for which  $\text{sim}(w, c_i) \geq \delta$

### Support for automatic tagging of HTML data:

- heuristic rules (`<homepage>`, `<publication>`, table headings, etc.)
- named-entity recognition (persons, companies, cities, temporal phrases)

### Query keyword disambiguation:

- contexts  $\text{con}(w)$  and  $\text{con}(c_i)$  for word  $w$  and concepts  $c_i \in \{c_1, \dots, c_k\}$
- word-bag similarity  $\text{sim}(\text{con}(w), \text{con}(c_i))$  based on cos or KL diff
- choose meaning  $\text{argmax}_{c_i} \{\text{sim}(\text{con}(w), \text{con}(c_i))\}$

### Mapping of concept-value query conditions onto Deep-Web portals:

~sheetmusic ~ „flute“ → instrument = (flute | piccolo | recorder)  
 → category = reeds  
 → style = ( classical | jazz | folk )

## Benefit from Ontology Service

Ontology service accessible via SOAP or RMI  
 Ontology filled with WordNet, geo gazetteer,  
 focused crawl results, extracted tables & forms

### Threshold-based

substitute ~w by

### Support for auto

- heuristic rules (
- named-entity re

### Query keyword a

- contexts  $\text{con}(w)$ ,
- word-bag similk
- choose meaning

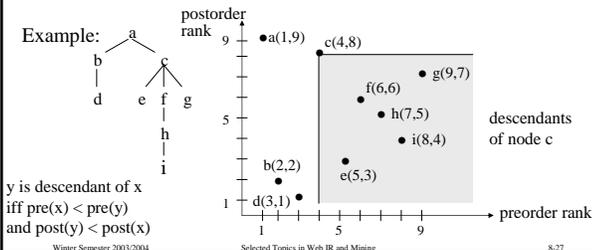
### Mapping of concept-value query conditions onto Deep-Web portals:

~sheetmusic ~ „flute“ → instrument = (flute | piccolo | recorder)  
 → category = reeds  
 → style = ( classical | jazz | folk )

## Index Structures for Path Expressions on Trees

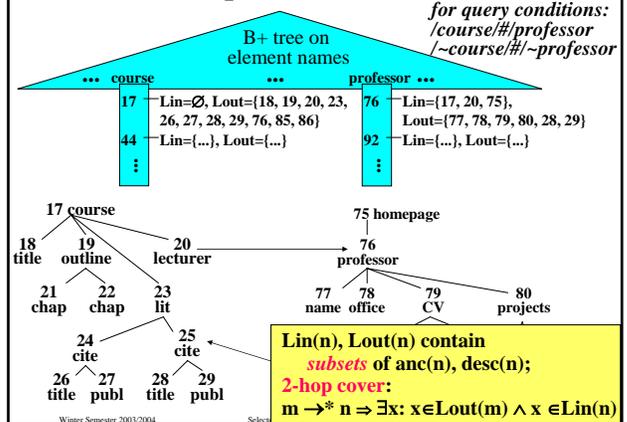
### Index based on Pre-/Postorder Numbering Scheme:

- Store with each element the
    - rank of the element in a preorder traversal of the doc tree and the
    - rank of the element in a postorder traversal.
  - Build multidimensional index over pre-/postorder ranks (and element ids, element names, and nesting depths)
- efficient eval. of XPath axes possible by appropriate window queries



## HOPi: 2-Hop-Cover based Path Index

for query conditions:  
 /course/#/professor  
 /~course/#/~professor



## Constructing a 2-Hop Cover

### Definition (E. Cohen et al., SODA 2002):

a 2-hop cover of a graph  $G=(V,E)$

is a labeling  $(\text{Lin}, \text{Lout})$  of all nodes where

1.  $\text{Lin}(n) \subseteq \{m \mid m \rightarrow^* n\}$ ,  $\text{Lout}(n) \subseteq \{p \mid n \rightarrow^* p\}$ , and
2.  $\forall (m,n) \in E \exists$  center node  $x$  with  $x \in \text{Lout}(m) \wedge x \in \text{Lin}(n)$

### Theorem (Cohen et al.):

The size of a 2-hop cover is  $\sum_{n \in V} |\text{Lin}(n)| + |\text{Lout}(n)|$ .

Finding a minimal 2-hop cover is NP-complete.

Polynomial Algorithm with  $O(\log V)$  updates  
 $T^* := E+ // \text{uncovered connections};$   
 while  $T^* \neq \emptyset$  {  
 + keep center graphs in priority queue  
 + incrementally update center subgraphs  
 + avoid complete transitive closure  
 + support incremental updates

for each node  $n$  construct center graph  
 $C(n) := \{(m,p) \mid (m,n), (n,p) \in E+\}$ ;  
 find node  $n$  with densest subgraph  $S(n)$  of  $C(n) \cap T^*$ ;  
 $\text{Lin}(n) := \text{sources of } S(n)$ ;  $\text{Lout}(n) := \text{sinks of } S(n)$ ;  
 remove edges of  $S(n)$  from  $T^*$ ;

## Efficient HOPi Construction

### Divide-and-conquer:

- Partition  $G$  by partitioning the XML document graph (using greedy heuristics) with node weights = #elems in doc & edge weights = #cross-doc links
- Compute 2-hop cover for each partition
- Merge covers:

for each cross-partition edge  $x \rightarrow y$  with  $x \in P, y \in Q$   
 add  $x$  to  $\text{Lout}(a)$  for all  $a \in P$  with  $a \rightarrow^* x$  and  
 to  $\text{Lin}(b)$  for all  $b \in P$  with  $y \rightarrow^* b$

### Implementation:

stores  $\text{Lin}$  and  $\text{Lout}$  sets in database tables  
 $\text{Lin}(\text{Id}, \text{InId})$  with indexes on  $\text{Id}$   $\text{InId}$  and  $\text{InId Id}$   
 $\text{Lout}(\text{Id}, \text{OutId})$  with indexes on  $\text{Id}$   $\text{OutId}$  and  $\text{OutId Id}$   
 Elems ( $\text{Id}, \text{ElemName}$ )  
 efficiently supports connection queries for all XPath axes  
 on arbitrary XML data graphs

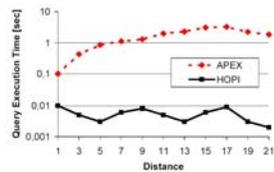
## Experimental Results for HOPI

for **DBLP data** in XML:  
 419 000 docs with 5 Mio. elems  
 and 63 000 links  
 306 Mio. connections (2.4 GB)

with 53 partitions, HOPI has  
 27 Mio. connections (415 MB)

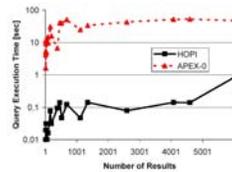
for query

//inproceedings[id=...]/author[id=...:]



for query

//inproceedings[id=...]/#



on **synthetic data** with Zipf-like indegrees and outdegrees  
 HOPI compresses TC by factor of 2 to 30

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.31

## Towards more Efficient Query Processing for XXL

interpret conditions of the form  
 ~course ~ „Internet“ and ~topic ~ „Performance“  
 as top-k queries with a score aggregation over  
 multiple index lists that come from  
 the ontology service and the content index

- apply Fagin-style top-k algorithms to retrieve best docs for similarity conditions
- then test exact-match conditions (needs pipelining for top-k evaluation)
- and identify result elements, paths, subgraphs

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.32

## 8.3 XSearch

Data:

set of XML trees  
 with interior nodes having names and leaf nodes having contents

Queries:

set of generalized keywords of the form

name:content, name:, :content,

referring to element names and contents,

with each condition optionally having a + flag for mandatory matches

Key idea:

results should be semantically coherent tree fragments

Definition:

element e satisfies condition n:c if

e has label n and a descendant whose contents contains c

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.33

## Example Data (1)

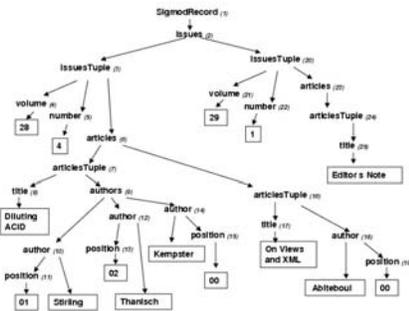


Figure 1: Part of the Sigmod Record document tree.

from: S. Cohen et al., XSearch: a Semantic Search Engine for XML, VLDB Conference, 2003

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.34

## Example Data (2)

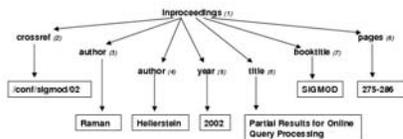


Figure 2: Part of the DBLP document tree.

from: S. Cohen et al., XSearch: a Semantic Search Engine for XML, VLDB Conference, 2003

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.35

## Query Answers

For two nodes  $n, n'$  in a tree the *interconnection tree*  $T_{n,n'}$  consists of  $\text{lca}(n, n')$  as the root and the paths from  $\text{lca}(n,n')$  to  $n$  and  $n'$ .

- Nodes  $n, n'$  are *meaningfully related*, denoted  $n \approx n'$ , if  $T_{n,n'}$
- does not contain two distinct nodes with the same name or
  - the only two distinct nodes with the same name are  $n$  and  $n'$

A set  $N$  of nodes is

*all-pairs related*, denoted  $\approx_s(N)$ , if  $n \approx n'$  for all  $n, n' \in N$  and

*star-related*, denoted  $\approx_s(N)$ , if there is  $n^* \in N$  s.t.  $n \approx n^*$  for all  $n \in N$ .

For a query with conditions  $c_1 \dots c_k$  the sequence  $n_1 \dots n_k$  of nodes and null values is an *all-pairs answer* if

- the non-null elements in  $\{n_1, \dots, n_k\}$  are all-pairs related,
- $n_i$  is not the null-value if  $c_i$  is a mandatory condition, and
- $n_i$  satisfies  $c_i$  if it is not the null value.

A *star-related answer* is analogously defined.

An answer  $N'$  for a query  $q$  subsumes answer  $N$  if  $N'$  is equal to  $N$  on all non-null elements.

$N$  is a *maximal answer* if every  $N'$  that subsumes  $N$  is equal to  $N$ .

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.36

## Query Answers

For two nodes  $n, n'$  in a tree the *interconnection tree*  $T_{n,n'}$  consists of  $\text{lca}(n, n')$  as the root and the paths from  $\text{lca}(n, n')$  to  $n$  and  $n'$ .

Nodes  $n, n'$  are *meaningfully related*, denoted  $n \approx n'$ , if  $T_{n,n'}$

- does not contain two distinct nodes with the same name or
- the only two distinct nodes with the same name are  $n$  and  $n'$

A set  $N$  of nodes is

*all-pairs related*, denoted  $\approx_i(N)$ , if  $n \approx n'$  for all  $n, n' \in N$  and *star-related*, denoted  $\approx_s(N)$ , if there is  $n^* \in N$  s.t.  $n \approx n^*$  for all  $n \in N$ .

For a query with conditions  $c_1 \dots c_k$  the sequence  $n_1 \dots n_k$  of nodes and null values is an *all-pairs answer* if

- the non-null elements in  $\{n_1, \dots, n_k\}$  are all-pairs related,
- $n_i$  is not the null-value if  $c_i$  is a mandatory condition, and
- $n_i$  satisfies  $c_i$  if it is not the null value.

A *star-related answer* is analogously defined.

An answer  $N'$  for a query  $q$  subsumes answer  $N$  if  $N'$  is equal to  $N$  on all non-null elements.

$N$  is a *maximal answer* if every  $N'$  that subsumes  $N$  is equal to  $N$ .

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.27

## Ranking Answers (1)

For query word  $w$  and leaf node  $n$  use  $\text{tf}^* \text{idf}$  as a weight of node  $n$ .

For query word  $w$  and interior node  $n$  use the sum of weights over all leaf nodes below  $n$ .

For label  $l$  and node  $n$  use  $l$  as a weight if  $n$ 's label is  $l$ , 0 otherwise.

Represent each node as an  $L * C$ -dimensional vector

with the above weights as components, where  $L$  is # of all possible labels and  $C$  the # of distinct words.

For query word  $w$  and label  $l$  set *query weight* to  $\text{tf}^* \text{idf}$  for condition  $l:w$ , 1.0 for condition  $:w$ , and user-specified (importance) weight for condition  $l$ : in all affected dimensions.

The *similarity score for answer set*  $N$  and query  $q$ ,  $\text{sim}(q, N)$ , is the sum, over all  $n \in N$ , of the cosine similarities between  $n$  and  $q$ .

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.38

## Ranking Answers (2)

For answer set  $N$  to query  $q$  define

$\text{tsize}(N) = \#$  nodes in interconnection tree of  $N$

$\text{ancdes}(N) = \#$  node pairs  $(n, n')$  in  $N$

where  $n$  is ancestor of  $n'$  or vice versa

The total score of answer  $N$  to query  $q$  is:

$$\frac{\text{sim}(q, N)^\alpha}{\text{tsize}(N)^\beta} \cdot (1 + \gamma \text{ancdes}(N))$$

with calibration parameters  $\alpha, \beta, \gamma$

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.39

## Interrelationship Indexing

Goal: efficiently testing nodes  $n, n'$  if they are meaningfully related

**Lemma:**

If  $n$  is ancestor of  $n'$ , then  $n \approx n'$  iff

$n \approx \text{parent}(n')$  and  $\text{label}(n) \neq \text{label}(\text{parent}(n'))$  and  $\text{child}(n) \approx n'$  and  $\text{label}(\text{child}(n)) \neq \text{label}(n')$ .

If neither  $n$  is ancestor of  $n'$  nor vice versa, then  $n \approx n'$  iff

$n \approx \text{parent}(n')$  and  $\text{label}(n) \neq \text{label}(\text{parent}(n'))$  and  $\text{parent}(n) \approx n'$  and  $\text{label}(\text{parent}(n)) \neq \text{label}(n')$ .

**Dynamic programming algorithm**

on Boolean matrix  $\text{interrel}[1..\#\text{nodes}]$  with depth-first node numbering:

```

for i := #nodes - 1 down to 0 { for j := i+1 to #nodes {
  if i is ancestor of j {
    let ch(i) be child of i on path to j and par(j) be parent of j;
    interrel[i,j] := interrel[ch(i),j] and label(ch(i)) != label(j) and
      interrel[i,par(j)] and label(i) != label(par(j)); } } };
for i := 1 to #nodes - 1 { for j := i+1 to #nodes {
  if i is not an ancestor of j {
    let par(i) be parent of i and par(j) be parent of j;
    interrel[i,j] := interrel[par(i),j] and label(par(i)) != label(j) and
      interrel[i,par(j)] and label(i) != label(par(j)); } } };
    
```

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.40

## Experimental Results

on DBLP sample + SIGMOD Record data

queries: 1)  $q_{kw} = \{+: \text{Buneman}, +: \text{database}\}$

2)  $q_{kw} = \{+: \text{author}, +: \text{title}\}$

3)  $q_{kw+tag} = \{+: \text{author}, +: \text{Buneman}, +: \text{title}, +: \text{database}\}$

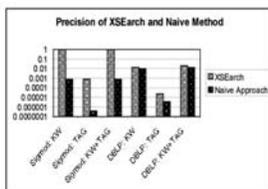


Figure 7: Precision of  $q_{kw}$ ,  $q_{kw+tag}$  and  $q_{kw+tag}$  for Sigmod Record and DBLP.

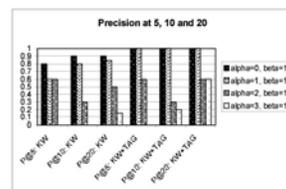


Figure 8: P@5, P@10 and P@20 of  $q_{kw}$  and  $q_{kw+tag}$  for DBLP.

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.41

## 8.4 XRank

Data: interlinked XML documents

Queries: simple keyword queries

Query results: ranked lists of elements

Key ideas:

- result ranking should consider
  - element-wise PageRank-style authorities
  - tree-node proximity of keyword-matching nodes
- query results are the most specific elements that have children with all keywords present

Winter Semester 2003/2004

Selected Topics in Web IR and Mining

8.42

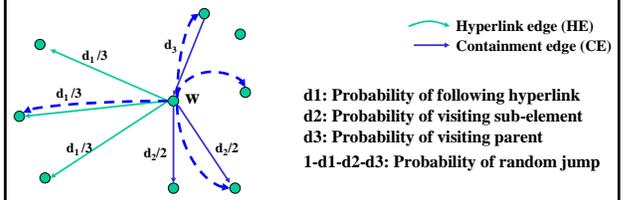
## Node Proximity Aware Scoring

For query keyword  $w$  consider element  $e$  such that  $e^t$  contains  $w$  and the path (within the tree) between  $e$  and  $e^t$  has length  $t$ . We define  $score(e, e^t, w) = score(e^t, w) * decay(e, e^t)$  with  $score(e^t, w) = ElemRank(e^t)$  if  $e^t$  contains  $w$ , 0 otherwise,  $decay(e, e^t) = \delta^{t-1}$ , and calibration parameter  $\delta, 0 < \delta < 1$ . If multiple  $e^t$  exist that contain  $w$  then  $score(e, w) = \max\{score(e, e^t, w)\}$ .

For query with keywords  $w_1, \dots, w_k$   
 $score(e, w_1, \dots, w_k) = (\sum_{i=1..k} score(e, w_i)) * prox(e, w_1, \dots, w_k)$   
 where  
 $prox(e, w_1, \dots, w_k) = size(smallest\ text\ window\ containing\ w_1, \dots, w_k)^{-1}$   
 measures the proximity of keywords in the linearized text below  $e$

## PageRank-Style Authority for Elements

*ElemRank*: define random walk on element graph and compute stationary authorities using standard power iteration

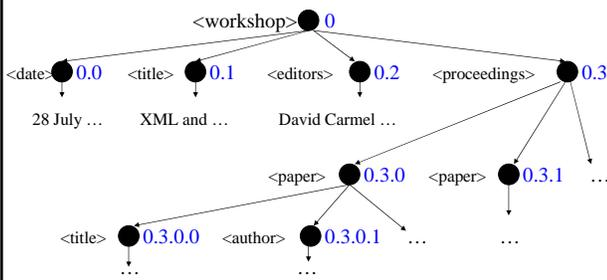


$$\mathcal{E}(v) = d_1 \sum_{(u,v) \in HE} \frac{\mathcal{E}(u)}{N_h(u)} + d_2 \sum_{(u,v) \in CE^+} \frac{\mathcal{E}(u)}{N_c(u)} + d_3 \sum_{(u,v) \in CE^-} \mathcal{E}(u) + \frac{1-d_1-d_2-d_3}{N_d \times N_{dc}(v)}$$

from: Lin Guo's presentation of Guo et al.: XRANK: Ranked Keyword Search over XML Documents, SIGMOD 2003

## Dewey Numbering Scheme

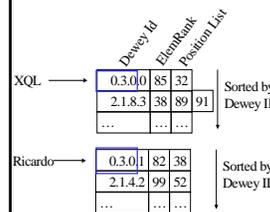
Dewey decimal numbering for elements:  
 ancestor's ID is prefix of descendant's ID



## Dewey Inverted Lists (DIL) Indexing

**Challenges for Query Processing:**

- Merge multiple inverted lists  
 - Not equality merge-join
- How to suppress spurious results  
 - Most specific results
- Capture the 2-dimensional proximity



Algorithm that addresses above issues in a **single scan** over inverted lists

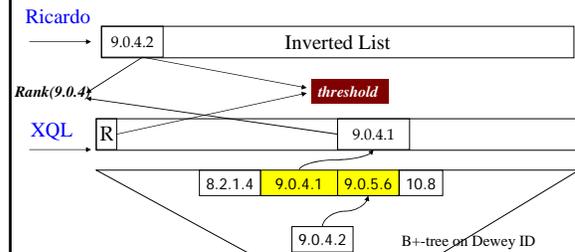
## Query Processing with DIL

- Dewey Inverted Lists (DIL) are sorted by Dewey ID order
- Goal: find top  $k$  results by a **single scan** over inverted lists
- Algorithm
  - Merge query keyword inverted lists in Dewey ID order
    - Ensures that entries with common prefixes are processed together
- Compute Longest Common Prefix (LCP) of Dewey IDs
  - LCP ensures most specific results

alternative: sort index lists by ElemRank  
 → **Ranked Dewey Inverted Lists (RDIL)**

## Query Processing with RDIL

1. How to find the most specific result?  
 Build and use B+ tree on Dewey IDs for each inverted list



2. When to stop scanning inverted lists?  
 Extend Fagin's TA:  
 threshold = aggregate score upper bounds with decay and prox set to 1.0.

### Literature

- Anja Theobald, Gerhard Weikum: Adding Relevance to XML, 3rd International Workshop on Web and Databases (WebDB), 2000, LNCS 1997, Springer.
- Ralf Schenkel, Anja Theobald, Gerhard Weikum: Ontology-enabled XML Search, in: H. Blanken et al. (Eds.), Intelligent Search on XML Data, LNCS 2818, Springer, 2003.
- Ralf Schenkel, Anja Theobald, Gerhard Weikum: HOPI: An Efficient Connection Index for Complex XML Document Collections, EDBT Conference, 2004.
- Sara Cohen, Jonathan Mamou, Yaron Kanza, Yehoshua Sagiv: XSEarch: A Semantic Search Engine for XML, VLDB Conf., 2003.
- Lin Guo, Feng Shao, Chavdar Botev, Jayavel Shanmugasundaram: XRank: Ranked Keyword Search over XML Documents, SIGMOD Conf., 2003.
- Sihem Amer-Yahia, SungRan Cho, Divesh Srivastava: Tree Pattern Relaxation, EDBT Conference, 2002.