

## LMU

#### Web Queries

#### From a Web of Data to a Semantic Web

François Bry, Tim Furche, Klara Weiand

based on work in the European project KiWi "Knowledge in a Wiki"









#### keyword search

#### ranked collection of documents





but what does Web querying do?









if you believe that I have some bank stocks for you ...



#### query: value of specific element

#### Share value: 27.40\$



	Web Coarob	Mah Quariaa
		web Queries
Scale	<b>"Web"</b> , TB/PBs	a few "documents", GBs
Parallelizability	very <b>high</b> (NC)	for basic selection lang. <b>high</b> , otherwise very <b>low</b>
Data	<b>independent</b> documents, heterogenous	trees (XML) or graphs (RDF), homogenous
Used by	(almost) everyone, many casual users	few experts
Expertise Level/Knowledge required	low	very <b>high</b>
Expressiveness	very <b>low</b>	very <b>high</b>
Result presentation	Ranking, clustering	programmed
Matching	often fuzzy or vague	only <b>precise</b> answers
Return value	Documents (or summaries thereof)	parts of trees or graphs
Actionability	very low	very <b>high</b>



#### Web Queries

#### Search + Query = Easy + Automation

#### Goals:

- make web querying accessible to casual users
  - e.g., to enable data aggregation/mashups by users
- allow precise queries over vastly heterogeneous data
- precise queries and rules critical to the (Social) Semantic Web
  - unless they are accessible no widespread adoption
- **Classification** of **approaches** to combining Web search & query
- enhance search: add data extraction / object search
- enhance querying: add keyword search / information retrieval
- keyword-based QLs for structured data: grounds-up redesign

## Search + Query

#### Approach 1: Enhance Search

"Peek" into web documents

**Extract** data items / Web objects from web documents

to provide more fine-grained answers

#### Examples

- Google (Squared)
- Google Rich Snippets
- Yahoo! Search Monkey

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Drooling Dog Bar B Q - Colfax, CA

★★★★☆ 15 reviews - Price range: \$\$ Drooling Dog has some really good BBQ. I had the pulled pork sandwich, .... Drooling Dog BBQ is a great place to stop at on your way up the hill to Tahoe ... www.yelp.com/biz/drooling-dog-bar-b-q-colfax - 75k - Cached - Similar pages

## Search + Query Approach 2: Enhance Querying

- Enhance existing (XML) web query languages
  - by adding information retrieval functionality
  - ranking, scoring, fuzzy matching
- Examples
- XQuery and XPath Full Text 1.0, W3C Cand. Recommendation

## Search + Query Approach 2: Enhance Querying

- Enhance existing (XML) web query languages
  - by adding information retrieval functionality
  - ranking, scoring, fuzzy matching

#### Examples

XQuery and XPath Full Text 1.0, W3C Cand. Recommendation

doc('bib.xml')/bib/book[title ftcontains 'programming']

🗩 for	<pre>\$b score \$s in doc('bib.xml')/bib/book</pre>
where	<pre>\$b ftcontains 'web' ftand 'query'</pre>
order by	<pre>\$s descending</pre>
return	<title> {\$b//title} </title>

## Search + Query Approach 3: Keyword Queries

- Keyword query for structured Web (XML and RDF) data
  - apply web **search paradigm** to querying tree and graph data
  - operates in the same setting as e.g. XQuery and SPARQL
    - few or single document, no Web scale
  - but: easier handling of very heterogeneous data
- querying of semi-structured data through easy-to-use interface

#### Ultimate goal:

Easy usage combined with enough power

→ to automate data processing tasks

## Search + Query Approach

Keyword query for str

- apply web **search p**
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K. Weiand, T. Furche, and F. Bry. Quo vadis, web queries? In Web4Web, 2008.

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#### Focus of this tutorial

#### Web Queries

#### Overview

- 1. Summary of **web query research** in the 00s
  - 1.1. XML
  - 1.2. RDF
- 2. Keyword query languages
  - 2.1. Motivation
  - 2.2. Classification
  - 2.3. Issues
- 3. KWQL
- 4. Discussion and Outlook

# Part 1 Web Queries

## Summary of XML & RDF Query Language Research



#### Tree Data & Tree Queries

Data-XPath-XQuery

```
<bib xmlns:dc="http://purl.org/dc/elements/1.1/">
   <article journal="Computer Journal" id="12">
2
      <dc:title>...Semantic Web...</dc:title>
      <vear>2005</vear>
4
      <authors>
       <author>
6
         <first>John</first> <last>Doe</last> </author>
       <author>
8
         <first>Mary</first> <last>Smith</last> </author>
     </authors>
10
   </article>
   <article journal="Web Journal">
12
      <dc:title>...Web...</dc:title>
     <year>2003</year>
14
     <authors>
       <author>
16
         <first>Peter</first> <last>Jones</last> </author>
       <author>
18
         <first>Sue</first> <last>Robinson</last> </author>
     </authors>
20
   </article>
22 </bib>
```



ordered tree (sometimes graph), mostly uniform edges



ordered tree (sometimes graph), mostly uniform edges



what's new about trees? didn't we try that before?



axis for navigation/selection in tree, context, horizontal axes for order

#### Intro in 5 Points

- Data: rooted, ordered, unranked, finite trees (no ID/IDREF resolution)
- Paths are sequences of steps (axis & test on properties of node)
- adorned with existential predicates (in []) to obtain tree queries
- some more advanced features: value joins, aggregation, ...
- Answers are sets of nodes from the input document
- no variables, no construction, no grouping/ordering

#### A Success Story

- Commercial success: One of the most successful QLs
- widely implemented and used, several W3C standards

#### Research success:

- designed with little concern for formal "beauty"
- but with few restrictions: turns out it hit a formal sweet spot
- Expressiveness: monadic datalog (datalog with only unary intensional predicates, i.e., answer predicates)
  - also: two-variable first-order logic (FO<sup>2</sup>)
- Polynomial complexity, linear for navigational XPath

## A Success Story (cont.)

**Completeness:** falls short of being first-order complete

- for first-order completeness a "conditional axis" (UNTIL operator) needed
- e.g., all a nodes reachable by a path of only b nodes
- but: partially justified by results on elimination of reverse axes
  - **reverse** axis: parent, ancestor, preceeding, ...
  - eliminating these axes possible in navigational XPath
    - though in few cases at exponential cost or resulting in introduction of expensive node-identity joins
    - ▶ we can **safely ignore** them in most cases
  - in conditional XPath this elimination is not possible

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First Order Complete XPath Dialect. PODS 2004.

D. Olteanu, H. Meuss, T. Furche, and F. Bry. **XPath: Looking Forward**. XMLDM @ EDBT, LNCS 2490, 2002.

> C. Ley and M. Benedikt. How big must complete XML query languages be? ICDT 2009.

## A Success Story (cont.)

#### Evaluation:

- naïve decomposition:
  - sequence of joins, descendant with closure over child relation
- better: structural joins for descendant (single lookup using tree labeling)
  - e.g., pre/post encoding
  - fits nicely with relational storage



twig joins: stack-based, holistic, not easily adapted to relational DBS

## XPath: Navigation in Trees A Success Story (cont.)



efficient evaluation

3


# -2

that's it? everyone happy?

```
let $auction := doc('auction.xml')
for $0 in $auction/open_auctions/open_auction
return
<corrupt id='{ $0/@id }'>
{ if (sum($0/(initial | bidder/increase)) = $0/current)
then text { 'no' }
else $auction//people/person[@id = $0//@person]/name }
</corrupt>
```

Adds an enormous set of features to XPath

- price: highly complex language, Turing-complete, very hard to implement
- here: just some highlights
- **Variables:** XPath plus variables  $\rightarrow$  no longer  $FO^2$ 
  - "an article that has the conference's chair as author"
  - not any more polynomial, but NP-/PSPACE-complete
- Construction: harmless if only at end of query
  - but: composition (i.e., querying of data constructed in the same query)
    - "find all papers written by the author with the most papers"
    - NEXPTIME-complete or in EXPSPACE (can not be captured by datalog)

#### **Recursion:** programmable

- with composition  $\rightarrow$  Turing complete
- i.e., like Prolog or datalog with value invention
- Sequences: rather than sets
- without composition little effect
- with composition: drastically harder optimization as iteration semantics of a query must be precisely observed



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2008

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'WD/CR/PR'

'REC'

item

'WD/CR/PR'

'WD/CR/PR'

'WD/CR/PR'





P. Boncz, T. Grust, M. van Keulen, S. Manegold, J. Rittinger, J. Teubner.
 MonetDB/XQuery: A Fast XQuery Processor Powered by a Relational Engine, SIGMOD 2006.

significantly harder than XPath

composition & recursion expensive operations

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yet significant steps toward fast implementations

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M. Benedikt and C. Koch. Interpreting Tree-to-Tree Queries. ICALP 2006.

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#### Graph Data for the Web





what's new about that? isn't that just a fancy repr. of a ternary relation?

#### RDF: Semantics for the Web?

#### What's new about RDF?

- **data model** of RDF very similar to relational
  - but: "unique" identifiers in form of URIs that can be shared on the Web
    - uniqueness only thanks to careful assignment practice (unlike GUIDs)
    - human-readable (unlike GUIDs)
- but: existential information in form of blank nodes
  - ▶ like named null values in SQL, also known as Codd tables
- but: implied information due to RDF/S semantics (and, thus, entailment)
  - ▶ e.g., class hierarchy and typing, domain and range of properties, axiomatic triples
- notion of redundancy-free or lean graph/answer

$$= \{\theta : \operatorname{dom}(\theta) = \operatorname{Vars}((s, p, o)) \land t\theta \in D\}$$

$$= \llbracket pattern_1 \rrbracket_{\operatorname{Subst}}^D \bowtie \llbracket pattern_2 \rrbracket_{\operatorname{Subst}}^D$$

$$= \llbracket pattern_1 \rrbracket_{\operatorname{Subst}}^D \cup \llbracket pattern_2 \rrbracket_{\operatorname{Subst}}^D$$

$$= \llbracket pattern_1 \rrbracket_{\operatorname{Subst}}^D \land \llbracket pattern_2 \rrbracket_{\operatorname{Subst}}^D$$

$$= \llbracket pattern_1 \rrbracket_{\operatorname{Subst}}^D \bowtie \llbracket pattern_2 \rrbracket_{\operatorname{Subst}}^D$$

$$= \{\theta \in \llbracket pattern \rrbracket_{\operatorname{Subst}}^D : \operatorname{Vars}(condition) \subset \operatorname{dom}(\theta) \land \llbracket condition \rrbracket_{\operatorname{Bool}}^D(\theta)\}$$

$$\begin{aligned} \|(\theta) &= \left[ condition_{1} \right] \right]_{Bool}^{D}(\theta) \land \left[ condition_{2} \right] \right]_{Bool}^{D}(\theta) \\ &= \left[ condition_{1} \right] \right]_{Bool}^{D}(\theta) \lor \left[ condition_{2} \right] \right]_{Bool}^{D}(\theta) \\ &= \neg \left[ condition \right] \right]_{Bool}^{D}(\theta) \\ &= \upsilon \theta \neq \mathbf{nil} \\ &= \upsilon \theta \in \mathbf{I} \\ &= \upsilon \theta \in \mathbf{I} \\ &= \upsilon \theta \in \mathbf{B} \\ &= \upsilon \theta = literal \\ &= \upsilon \theta = \upsilon \theta \land \upsilon \theta \neq \mathbf{nil} \end{aligned}$$

 $\begin{bmatrix} triple \end{bmatrix}_{\text{Graph}}^{D} (\theta) = triple\theta \text{ if } \forall v \in \text{Vars}(triple) : v\theta \neq \text{nil}, \top \text{ otherwise}$   $\begin{bmatrix} template_1 \text{ AND } template_2 \end{bmatrix}_{\text{Graph}}^{D} (\theta) = \begin{bmatrix} template_1 \end{bmatrix}_{\text{Graph}}^{D} (\theta) \cup \begin{bmatrix} template_2 \end{bmatrix}_{\text{Graph}}^{D} (\theta)$   $\begin{bmatrix} \text{CONSTRUCT } t \text{ WHERE } p \end{bmatrix}^{D} = \bigcup_{\theta \in \llbracket P \rrbracket_{\text{Subst}}}^{D} \llbracket \text{std}(t) \rrbracket_{\text{Graph}}^{D} (\theta)$   $\begin{bmatrix} \text{SELECT } V \text{ WHERE } p \end{bmatrix}^{D} = \pi_V(\llbracket P \rrbracket_{\text{Subst}}^{D})$ 

 $\begin{bmatrix} (s, p, o) \end{bmatrix}_{\text{Subst}}^{D} \\ \begin{bmatrix} pattern_1 \text{ AND } pattern_2 \end{bmatrix}_{\text{Subst}}^{D} \\ \begin{bmatrix} pattern_1 \text{ UNION } pattern_2 \end{bmatrix}_{\text{Subst}}^{D} \\ \begin{bmatrix} pattern_1 \text{ MINUS } pattern_2 \end{bmatrix}_{\text{Subst}}^{D} \\ \begin{bmatrix} pattern_1 \text{ OPT } pattern_2 \end{bmatrix}_{\text{Subst}}^{D} \\ \begin{bmatrix} pattern \text{ FILTER } condition \end{bmatrix}_{\text{Subst}}^{D} \end{bmatrix}$ 

#### RDF: Semantics for the Web?

#### Simple RDF QL?

- SPARQL: Simple Protocol and RDF Query Language
- really simple?
- same expressiveness as full relational algebra, PSPACE-complete
- but: **no composition**, **no order**  $\rightarrow$  simpler than full SQL or XQuery
- really an **RDF** query language?
- blank node construction only limited (no quantifier alternation)
- talks vaguely about extension to entailment regimes
  - but no support in SPARQL as defined
- no support for lean answers (justified partially by high computational cost)

### RDF: Semantics for the Web? SPARQL: Simple RDF QL?

#### Complexity:

- SPARQL with only AND, FILTER, and UNION: NP-complete
  - ▶ i.e., no computationally interesting subset of **full** relational SPJU queries
- ► full SPARQL: PSPACE-complete
  - again no computationally interesting subset of full first-order queries
  - why? due to negation "hidden" in OPTIONAL
    - reduction from 3SAT using **isBound** to encode negation
- lacks completeness w.r.t. RDF transformations:
  - relational algebra: can express all PSPACE-transformations on relations
  - ► SPARQL: fails at the same for RDF

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J. Perez, M. Arenas, and C. Gutierrez. **Semantics and Complexity of SPARQL**. ISWC 2006 Select all persons and return **one** blank node connected to all these persons using "member"



#### **Not expressible** in SPARQL Why not? no **quantifier alternation**

SPA

limitations of blank node construction

Select all persons and return **one** blank node connected to all these persons using "member"



**Not expressible** in SPARQL Why not? no **quantifier alternation**  SPARQL

limitations of blank node construction

#### RDF: Semantics for the Web?

#### SPARQL: Future

- doesn't hit a sweet spot (like XPath)
  - at least from a formal, database perspective
    - contributions from database community very rare
- yet a significant improvement over most previous RDF QLs
- already forms basis for future QL research on RDF
- rule extensions for RDF
  - "From SPARQL to Rules" (Polleres, WWW 2007), Networked Graphs (Schenk et al, WWW 2008)
  - no or limited blank node support
  - ▶ but: with rules, restricted quantification as in SPARQL (∃∀) as expressive as unrestricted quantifier alternation

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F. Bry, T. Furche, C. Ley, B. Linse, B. Marnette, and O. Poppe. **SPARQLog: SPARQL with Rules and Quantification**. In: *Semantic Web Information Management: A Model-based Perspective*, 2009.

# RDF: Semantics for the Web? SPARQL: Summary

syntax for first-order queries on RDF

foundation for research but little innovation in the language itself

2

3

#### lackluster support for RDF specifics



# Part 2 Keyword Queries

#### Classification & main issues



#### Queries as Keywords

#### **Main Characteristics**

- Queries are (mostly) unstructured bags of words
- Used in general purpose web search engines
  - but also elsewhere (Amazon, Facebook,...)
- Implicit conjunctive semantics (with limitations)
- Often combined with IR techniques
  - ranking, fuzzy matching
- Research focuses on
- application to semi-structured data
- general structured data (Web objects & tables, relations)

#### Queries as Keywords

#### Why Keyword Qs for Structured Data

- Success in other areas
- Users are already familiarized with the paradigm
- Allow casual users to query structured data without having deep knowledge of
  - the query language
  - the structure & schema of the data
- Enable querying of heterogeneous data

# Queries as Keywords Classification of Keyword QLs

#### Data type

- ► XML
- ► RDF

#### Implementation

- stand-alone systems
- translation to conventional query language
- keyword-enhanced query languages

#### Queries as Keywords Classification of Keyword QLs

#### Complexity of atomic queries

- keyword-only
- Iabel-keyword
- keyword-enhanced
- Querying of elements
  - Values
  - Node labels
  - Edge labels

#### Queries as Keywords

#### XML Keyword QLs

	Stand-alone	Enhancement
Keyword-only	9	0
Label-keyword	2	0
Keyword-enhanced		3

#### Queries as Keywords

K. Weiand, T. Furche, and F. Bry. **Quo vadis, web queries?** In Web4Web, 2008.

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## Queries as Keywords

## **RDF** Keyword QLs

	Stand-alone	Translation
Keyword-only	2	2
Label-keyword	0	1
Keyword-enhanced		<b></b>

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## **RDF** Keyword QLs

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	Keyword-only	2	2
	Label-keyword	0	1
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## Queries as Keywords XML Keyword QLs More Popular

#### Time

first XML keyword query languages are as old as RDF

- Familiarity with XML
- Complexity of RDF
  - Graph-shaped
  - Labeled Edges
  - Blank nodes

## Queries as Keywords XML Keyword QLs More Popular

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### Queries as Keywords

### Focus **Issues**

- 1. Grouping keyword matches
- 2. Determining answer representations
- 3. Expressive power
- 4. Ranking
- 5. Limitations

# Computing Query Answers

# Turning keyword matches into **answers**

#### Computing Query Answers

## **Problem** Setting

- Query K yields match lists L<sub>1</sub> ... L<sub>|K|</sub>
- Answer sets S<sub>1</sub> ... S<sub>m</sub> are constructed from L
  - may contain either
    - ▶ only one match per match list (m = |K|) or
    - ► several ( $m \ge |K|$ )

Data-centric vs. document-centric XML



### Computing Query Answers

## **Problem** Setting

#### Entire XML document

- usually too big to serve as a good query answer
- Matched nodes alone
  - usually not informative
- Meaningful results
  - a smaller unit has to be found

#### Computing Query Answers

## Approaches

- 1. Determine entities based on the schema
  - only keyword matches within one entity or
  - apply LCA at entity-level
  - done manually or using schema partitioning with cardinality as criterion
    - ► Lowest Entity Node, Minimal Information Unit, XSeek...
- 2. Determine entities by connecting keyword matches
  - Answer entity:
    - contains (at least) one match for each keyword

## Computing Query Answers Connecting Keyword Matches

- Classic concept: Lowest Common Ancestor (LCA)
  Use LCA to find the maximally specific concept that is common to the keyword matches
- LCA: Lowest node that is ancestor to all elements in an answer set





# 2.1.1

# Improving LCA

# Approaches to **improve LCA** for computing answer entities

Computing Query Answers Lowest Common Ancestor (LCA)

- Many approaches to improve over LCA
  SLCA, MLCA, Interconnection Semantics, VLCA, Amoeba Join...
- Filter LCAs to avoid false positives
- Result is subset of LCA result

## Overview of Approaches

- 1. Interconnection relationship
- 2. XKSearch: Smallest LCA
- 3. Meaningful LCA (MLCA)
- 4. Amoeba Join
- 5. (Valuable LCA (VLCA), Compact LCA, CVLCA)
- 6. (Relaxed Tightest Fragment (RTF))
- 7. XRank: Exclusive LCA

- Idea: Two different nodes with the same label correspond to different entities of the same type.
- Two nodes are interconnected if,
  - on the shortest path between them,
  - every node label occurs only once
- Interconnection in answer sets:
  star-related: a node in S<sub>i</sub> interconnected with all nodes in S<sub>i</sub>
  all-pairs related: all nodes in S<sub>i</sub> pair-wise interconnected

## Interconnection Relationship

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S. Cohen, J. Mamou, Y. Kanza, and Y. Sagiv. XSearch: A Semantic Search Engine for XML. VLDB 2003

## Improving LCA Interconnection Relationship





## Interconnection Relationship

For |K| = 2, star-related = all-pairs related

• Consider  $S_1 = \{3, 8, 11\}$ 

## Interconnection Relationship

#### 3 and 8 are interconnected







## Interconnection Relationship

- S1={3, 8, 11} is
  - star-related but
  - not all-pairs related

False negative when using all-pairs relatedness

Consider S1={3,7,9}







- S<sub>1</sub>= $\{3,7,9\}$  is a false negative in both measures
- False positives for both when node labels are different but refer to similar concepts
  - e.g. "article" vs. "book" vs. "text"
- False negatives when node labels are identical but refer to different concepts
  - e.g. the name of a person vs. the name of a journal

## **XKSearch:** Smallest LCA (SLCA)

- Idea: Enhance LCA with a minimality constraint
- SLCA nodes are LCAs which
  - do not have LCA nodes among their descendants
- Similar to XRank/ELCA but stricter

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Y. Xu and Y. Papakonstantinou. Efficient Keyword Search for Smallest LCAs in XML Databases. SIGMOD 2005

## Improving LCA XKSearch: Smallest LCA (SLCA)



## Improving LCA XKSearch: Smallest LCA (SLCA)

No false positive





## SLCA: False Negatives



## Meaningful LCA (MLCA)

- MLCA: LCA where
  - ▶ for each pair of nodes,
  - no descendant LCA of nodes with the same label exists
- Similar to SLCA but less strict since only applies when node label constraint is fulfilled
- Problems similar to SLCA and Interconnection Semantics
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Y. Li, C. Yu, and H. V. Jagadish. Schema-Free XQuery. VLDB 2004





#### Amoeba Join

- S<sub>i</sub> is a valid answer set if  $LCA(S_i) \in S_i$ 
  - only allows matches where
    - one matched node is
    - an ancestor of (or identical to) all matched nodes
- K={Smith, web}, L<sub>1</sub>={11}, L<sub>2</sub>={3,23}, S<sub>1</sub>={11,3}, S<sub>2</sub>= {11,23}

#### Amoeba Join

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  - only allows matches where
    - one matched node is
    - ► an ancestor of (or identical to) all matched nodes
- K={Smith, web}, L<sub>1</sub>={11}, L<sub>2</sub>={3,23}, S<sub>1</sub>={11,3}, S<sub>2</sub>= {11,23}

T. Saito and S. Morishita. Amoeba Join: Overcoming Structural Fluctuations in XML Data. WebDB

#### Amoeba Join

#### K={web, Smith} — No false positive



#### Amoeba Join

#### K={web, Smith} — False negative



#### Amoeba Join

K={web, Smith, article}



#### Amoeba Join

#### K={web, Smith, article} - False positive



### Compact LCA (CLCA)

CLCA: LCA node of an answer set S<sub>i</sub>

- where LCA(S<sub>i</sub>) dominates all nodes in S<sub>i</sub>
- Node  $v_i$  **dominates** node  $v_j$  if there is no  $v_j \in S_i$  where LCA (S<sub>i</sub>) is descendant of  $v_i$
- Simpler: CLCA is an LCA
  - where no node in the associated answer set can be
    - ► part of a more specific answer set
- Similar to SLCA

#### **Compact Valuable** LCA (VLCA)

- CVLCA: CLCA node which is also a VLCA node
- Recall problems of SLCA and Interconnection Semantics

## **Relaxed Tightest Fragment** (RTF)

- Subtrees are complete with respect to matches
   while being as small as possible
- Similar to **XRank**:

maximum match without contained descendant LCAs

K={Smith, web}, L<sub>1</sub>={11}, L<sub>2</sub>={3,23}, S<sub>1</sub>={11,3,23}, S<sub>2</sub>= {11,3}, S<sub>3</sub>={11,23} and LCA(11,3)=2, LCA(11,14)=1, LCA(11,3,23)=1







#### **XRank:** Exclusive LCA

Idea: Prefer more specific LCAs
unless reason not to (more matches)
Result node is LCA node which
does not contain further LCAs or
which is still LCA if LCA subtrees are ignored
As before, K={Smith, web}, L<sub>1</sub>={11}, L<sub>2</sub>={3,23}, S<sub>1</sub>= {11,3}, S<sub>2</sub>={11,23} and LCA(11,3)=2, LCA(11,23)=1

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Lin Guo et al. XRANK: Ranked keyword search over XML documents. SIGMOD 2003

#### XRank: Exclusive LCA

#### Only (2) is a valid result node





#### XRank: Exclusive LCA

Objection:

XRank targeted at document-centric XML

- Does this solve the problem?
- Consider K={XML,RDF}

# XRank: Exclusive LCA $S_1=\{13\}, S_2=\{38,80\}...$ Is 10 a good return node?



#### Summary

- No heuristic with perfect precision and recall
- Data-driven solutions, not universally applicable
- Monotonicity and consistency are desirable but often violated
- In some cases, no heuristic produces suitable results

Summary

#### K={Smith,2003}



#### Summary

#### K={Doe, Smith} (1) bib (2) article (13) article (3) title (4) year (12) journal (16) authors (23) journal (5) authors (14) title (15) year (17) author (20) author (6) author (9) author 2005 **Computer Journal** ...XML... Web Journal ...Semantic Web... 2003 (19) last (21) first (10) first (11) last (18) first (22) last (7) first (8) last John Sue Robinson Doe Mary Peter Jones

#### Queries as Keywords

#### RDF

- Summarize RDF graph (Q2RDF, Q2Semantic)
- Generate queries or query results by connecting matches
  - Dijkstra's Algorithm (Q2RDF)
  - Kruskal's Minimum Spanning Tree Algorithm (SPARK)
  - Templates based on types (SemSearch)
  - Cost-based heuristics (Q2Semantic)



# Determining Return Values

# From an answer node to an **answer representation**

# Determining Return Values Approach 1: LCA

LCA (or similar) node, e.g. ELCA, MLCA





# Determining Return Values Approach 3: LCA & Path

Path from LCA to matched nodes

▶ e.g. BANKS, RTF





# Determining Return Values Comparison & Summary

- Neither approach is always satisfying
  - subtree may be too big
  - path or node may not provide enough information
- More controlled return value is desirable
  - use query elements to determine a suitable return value automatically

#### Determining Return Values

#### Exemplar: XSeek

- Processing steps:
  - 1. Match query on node labels and content
  - 2. Group matches using SLCA
  - 3. Extract **return nodes** from query: If a term in K matches a node label and no descendant content is matched, consider the term a return node (explicit)
  - 4. When there are no return nodes, return **SLCA entity subtree** (implicit)

#### Determining Return Values

#### Exemplar: XSeek

- Terms in the query that are not return nodes are search predicates i.e. keywords to find
- Entities and their attributes inferred from the schema using cardinality information

#### Final return value:

explicit or implicit return nodes and entities' attributes

#### Determining Return Values

#### Exemplar: XSeek








# Expressiveness

- Queries: unordered lists with implicit conjunction
   + Conjunction, disjunction (Multiway, Abbaci et al.)
  - + Inclusion, sibling, negation, precedence (Abbaci et al.)
- User-selected return value (XSeek, MIU)
- Numeric comparison operators (MIU)
- Optional terms (Interconnection)
- Iabel:keyword terms (Interconnection, XSearch)
- Keyword-enhanced languages

#### Expressiveness



# Expressiveness

- Transform query into binary tree
- Construct set of matching nodes and their ancestors for each term
- Bottom-up processing:
  - Apply operator to sets of nodes (i.e. intersection for conjunction)
  - The nodes remaining at the root are valid answers (LCA-like)



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F. Abbaci et al. Index and Search XML Documents by Combining Content. International Conference on Internet Computing 2006

# Ranking

- Size of answer subtree
- Distance between matched nodes and answer tree root node
- tf-idf/vector space model
- XRank: PageRank-like ranking

# XRank – Ranking factors

- Result specificity: vertical distance
- Keyword proximity: horizontal distance
- Hyperlink awareness: PageRank value, adapted for XML
  - distinction between XML edges and IDREF links
  - Bidirectional propagation between XML edges
  - distinction between following forward and backward links

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Lin Guo et al. XRANK: Ranked keyword search over XML documents. SIGMOD 2003

# Limitations

#### Mostly limited to tree data

- Determining semantic entities in structured data
  - Assumption: No element outside of LCA subtree is relevant
  - No universal solution, data-driven
- Relatively low expressiveness

# Limitations

#### Query answers

- Little control over selection
- Result may be too verbose or not informative enough
- No construction or aggregation
- Exception: Keyword-enhanced languages
- No querying of data in mixed formats (RDF & XML)



# Keyword-based query language for Semantic Wikis

# Semantic Wikis:

# The (Semantic) Web in the Small

- As on the Semantic Web we have
  - pages and links between them and annotations
  - content created by many different people
  - But we also have
  - central control, organization and administration
  - ► a small (or at least manageable) number of pages
  - Strong social factors and collaboration

→Wikis as a testbed for the "real" web

# KWQL: Keyword-Based QL for Wikis Characteristics

- KWQL can access all elements the user interacts with
  - Combined querying of text, annotation and metadata
  - Querying of informal to formal annotations
  - Combination of selection criteria from several data sources in one query
- Aggregation and construction
   Data construction
  - Embedded queries
  - Continuous queries

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F. Bry and K. Weiand. Flavors of KWQL, a Keyword Query Language for a Semantic Wiki. SOFSEM, 2010.

# **Characteristics**

- Varying complexity of queries
  - Simple label-keyword queries
  - Conjunction/disjunction/optional
  - Structural queries
  - Link traversal



# visKWQL

- KWQL's visual counterpart
- Query by example paradigm
- Round-tripping between KWQL and visKWQL
  - Visualization of textual queries
  - visKWQL as a tool to learn KWQL

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# Part 4 Conclusion

# Web Queries

# Conclusion

- Is it even possible to find a universal grouping mechanism?
- How easy to use can a query language be while still being powerful enough?
- How complicated can a query language be without becoming too hard to use for casual users?

### Web Queries

# Conclusion

# Slides and links at <a href="http://pms.ifi.lmu.de/wise">http://pms.ifi.lmu.de/wise</a>