O NICTA

Querying and Exchanging XML and RDF on the Web

WWW'12 Tutorial

Sherif Sakr (NICTA and UNSW) Axel Polleres (Siemens AG)



Australian Government Department of Broadband, Communications and the Digital Economy

Australian Research Council







THE UNIVERSITY OF QUEENSLAND

NICTA Copyright 2012

From imagination to impact

NICTA Funding and Supporting Members and Partners

Queensland

Tutorial Overview

Session 1



XQuery Overview – Sherif

SPARQL Overview – Axel XSPARQL: a combined language – Axel

Session 2 XQuery implementations – Sherif SPARQL implementations – Sherif XSPARQL implementations – Axel (optional) Compression formats for XML+RDF: EXI+HDT – Sherif

Q/A - Discussion









Australian Government Department of Broadband, Communications and the Digital Economy

Australian Research Council



Queensland Government





QUT





OF QUEENSLAND

THE UNIVERSITY OF SYDNEY

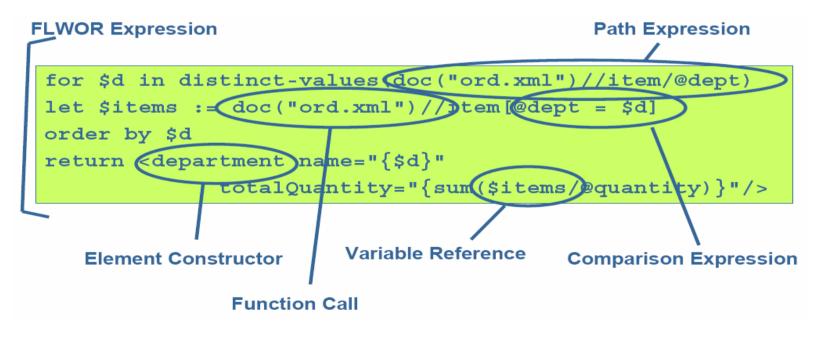




XQuery is a declarative language in which a query is represented as an

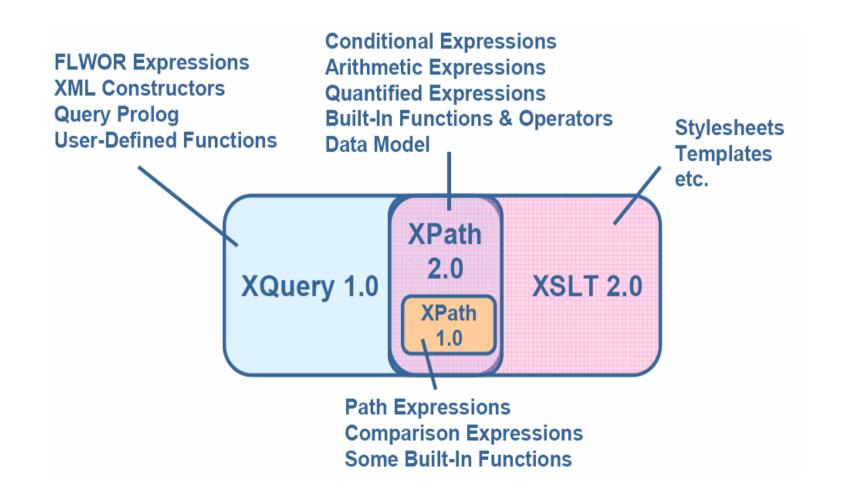
expression.

• XQuery expressions can be nested with full generality.



XQuery, XSLT and XPath







- The XQuery language is designed to operate over ordered, *finite sequences of items* as its principal data type.
- The evaluation of any XQuery expression yields an ordered sequence of n >= 0 items.
- These items can be:
 - Atomic values (integers, strings, ..., etc)
 - Unranked XML tree nodes.

Items and Ordered Sequences



 A sequence of *n* items X_i is written in parentheses and commaseparated form

$(X_1, X_2, ..., X_n)$

- A single item X and the singleton sequence (X) are equivalent.
- Sequences *can not* contain other sequences (nested sequences are implicitly flattened)

```
(0, (), (1, 2), (3)) = (0, 1, 2, 3)
(0,1) \neq (1,0)
```

- Sequences *can* contain *duplicates* (0, 1, 1, 2)
- Sequences may be *heterogeneous*

(42, "foo", 4.2, <a>)



- Programming language features:
 - Explicit iteration and variable bindings (for, let, ...).
 - Recursive, user-defined functions.
 - Regular expressions, strong [static] typing.
 - Ordered sequences (much like lists or arrays).
- Query language features:
 - Filtering.
 - Grouping.
 - Joins.

XQuery



- A strongly-typed, Turing-complete XML manipulation language
 - Attempts to do static type checking against XML Schema
 - Based on an object model derived from Schema
- Unlike SQL, fully compositional, highly orthogonal:
 - Inputs & outputs collections (sequences or bags) of XML nodes
 - Anywhere a particular type of object may be used, may use the results of a query of the same type
 - Designed mostly by DB and functional language people
- Attempts to satisfy the needs of data management and document management
 - The database-style core is mostly complete (even has support for NULLs in XML!!)
 - The document keyword querying features are still in the works shows in the order-preserving default model



- Extracting information from a database for use in web service.
- Generating summary reports on data stored in XML database.
- Searching textual documents on the web for relevant information.
- Transforming XML data to XHTML format to be published on the web.
- Pulling data from different databases to be used for application integration.
- Splitting up an XML document into multiple XML documents.

XQuery Expressions



- Path expressions.
- FLWOR expressions.
- Expressions involving operators and functions.
- Conditional expressions.
- Quantified expressions.
- List constructors.
- Element constructors.
- Expressions that test or modify datatypes



- In a sense, the *traversal* or *navigation* of trees of XML nodes lies at the core of every XML query language.
- XQuery embeds **XPath** as its tree navigation sub-language.
- Every XPathexpression is a correct XQuery expression.
- Since navigation expressions extract (potentially huge volumes of) nodes from input XML documents, efficient XPath implementation is a prime concern to any implementation of an XQuery processors.



Y Each path consists of one or more **steps**, syntactically separated by /

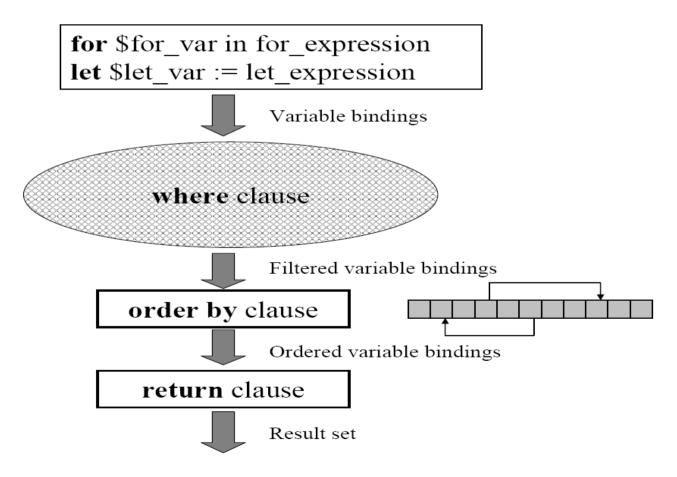
 $s_0/s_1/\ldots/s_n$

Y Each step acts like an operator that, given a sequence of nodes (the context set), evaluates to a sequence of nodes.
Y XPath defines the result of each path expression to be duplicate free and sorted in document order.





 A FLWOR expression binds some expressions, applies a predicate, and constructs a new ordered result.



FLWOR Expression



Has an analogous form to SQL's

SELECT. FROM. WHERE. GROUPBY. ORDER BY

- The model: bind nodes (or node sets) to variables; operate over each legal combination of bindings; produce a set of nodes
- "FLWOR" statement:

for {iterators that bind variables}

let {collections}

where {conditions}

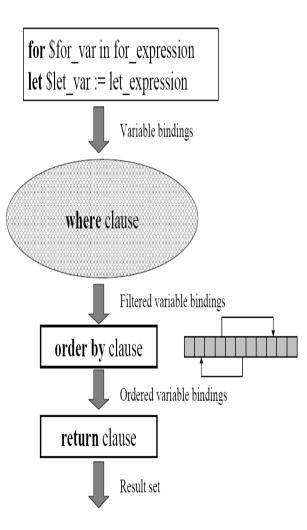
order by {order-conditions}

return {output constructor}

FLWOR Expression



- The for construct successively binds each item of an expression (expr) to a variable (var), generating a so-called tuple stream.
- This tuple stream is then **filtered** by the where clause, retaining some tuples and discarding others.
- The return clause is evaluated once for every tuple still in the stream.
- The result of the expression is an ordered sequence containing the concatenated results of these evaluations.



FLWOR Expression



Iteration for \$x in (3,2,1) \Rightarrow (3, "*", 2, "*", 1, "*") return (\$x,"*") for \$x in (3,2,1) \Rightarrow (3,2,1,"*") return \$x,"*" for \$x in (3,2,1) (3,"a",3,"b", return for y in ("a", "b") \Rightarrow 2, "a", 2, "b", return (\$x,\$y) 1,"a",1,"b")



for \$book in document("bib.xml")//book,
 \$quote in document("quotes.xml")//listing
where \$book/isbn = \$quote/isbn
return
<book>
 { \$book/title }
 { \$puote/price }
</book>



```
for $book in document("bib.xml")//book
return
<book>
      { $book/title }
for $review in document("reviews.xml")//review
where $book/isbn = $review/isbn
return $review/rating
</book>
```



- for iterates on a sequence, binds a variable to each node.
- let binds a variable to a sequence as a whole.
- Together, they are used for representing aggregation and grouping expressions.



- Infix and prefix operators (+, -, *,...).
- Parenthesized expressions.
- Arithmetic and logical operators.
- Collection operators UNION, INTERSECT and EXCEPT.
- Infix operators BEFORE and AFTER (<< , >>).
- User functions can be defined in XQuery.



	comparison
NUUG	comparison

Node comparisonsbased on identity and document order: e_1 is e_2 nodes $e_{1,2}$ identical? $e_1 << e_2$ node e_1 before e_2 ? $e_1 >> e_2$ node e_1 after e_2 ?

Node comparison examples



Value	comparing single values	eq, ne, It, Ie, gt, ge
	Untyped data => string	
General	Existential quantification	=, !=, <=, <, >, >=
	Untyped data => coerced to other operand's type	
Node	for testing identity of single nodes	is, isnot
Order	testing relative position of one node vs. another (in document order)	<<, >>





- Logical Operators "and" and "or".
- The concept of Effective Boolean Value(EBV) is key to evaluating logical expressions.
 - **EBV** of an empty sequence is *false*.
 - **EBV** of a non-empty sequence containing only nodes is *true*.
 - EBV is the value of the expression if the expression evaluates to a value of type xs:boolean.
 - **EBV** is an error in every other case.
- e.g: The expression "() and true()" evaluates to false(since () is false)

XQuery: Built-in Functions



- Over 100 functions built into XQuery.
- String-related
 - substring, contains, concat,...
- Date-related
 - current-date, month-from-date,...
- Number-related
 - round, avg, sum, …
- Sequence-related
 - index-of, distinct-values,...
- Node-related
 - data, empty,...
- Document-related
 - doc, collection, ...
- Error Handling
 - error, exactly-one, …
-

XQuery: User-Defined Functions



- XQuery expressions can contain user-defined functions which encapsulate query details.
- User-defined functions may be collected into modules and then 'import'ed by a query.

Declaration of *n***-ary function** *f* **with body** *e*

declare function $f(p_1 \text{ as } t_1, \ldots, p_n \text{ as } t_n)$ as $t_0 \{ e \}$

- ▷ If t_i is omitted, it defaults to item()*.
- \triangleright The pair (f, n) is required to be unique (overloading).
- \triangleright Atomization is applied to the *i*-th parameter if t_i is atomic.



Reverse a sequence

```
Reversing a sequence does not inspect the sequence's items in any way:
```

```
declare function reverse($seq)
{ for $i at $p in $seq
   order by $p descending
   return $i
};
```

```
reverse((42,"a",<b/>,doc("foo.xml")))
```

Conditional Expression



- Syntax:

if (expr1) then expr2 else expr3

- if EBV of *expr1* is true, the conditional expression evaluates to the value of *expr2*, else it evaluates to the value of *expr3*.
- Parentheses around if expression (*expr1*) are required.
- else is always required but it can be just else ().
- Useful when structure of information returned depends on a condition.
- Can be nested and used anywhere a value is expected. if(\$book/@year <1980) then<old>{\$x/title}</old> else<new>{\$x/title}</new>



• Used as an alternative way of writing the FLWOR expressions.

FLWOR:for \$a in document("bib.xml")//article where \$a/year < 1996 return \$a

```
Conditional:for $a in document("bib.xml")//article
return
If ($a/year < 1996)
then$a
else()
```

Quantified Expressions



• Syntax:

[some | every]\$var in expr satisfies test_expr

- Quantified expressions evaluate to a boolean value.
- Evaluation:
 - \$var is bound to each of the items in the sequence resulting from expr.
 - For each binding, the *test_expr* is evaluated.
 - In case of
 - Existential quantification ("some"), if at least one evaluation of test_expr evaluates "true", the entire expression evaluates "true".
 - Universal quantification ("every"), all evaluations of <u>test_exp</u> rmust result in an EBV of "true" for the expression to return "true".



- Existential Quantification
 - Give me all books where "Sailing" appear at least once in the same paragraph.

for \$b in document("bib.xml")//book
where some \$p in \$b//para satisfies(contains(\$p,"Sailing"))
return \$b/title



- Universal Quantification
 - Give me all books where "Sailing" appears in every paragraph.

for \$b in document("bib.xml")//book
where every \$p in \$b//para satisfies(contains(\$p,"Sailing"))
return \$b/title



A list may be constructed by enclosing zero or more

expressions in square brackets, separated by commas.

For example, [\$x, \$y, \$z] denotes a list containing three

members represented by variables.

• [] denotes an empty list.



XML tree fragments may be constructed "on the fly" in queries:

for \$i in doc("auction.xml") /site/regions/australia/item

return

<item>

<name>{\$i/name/text()}</name>

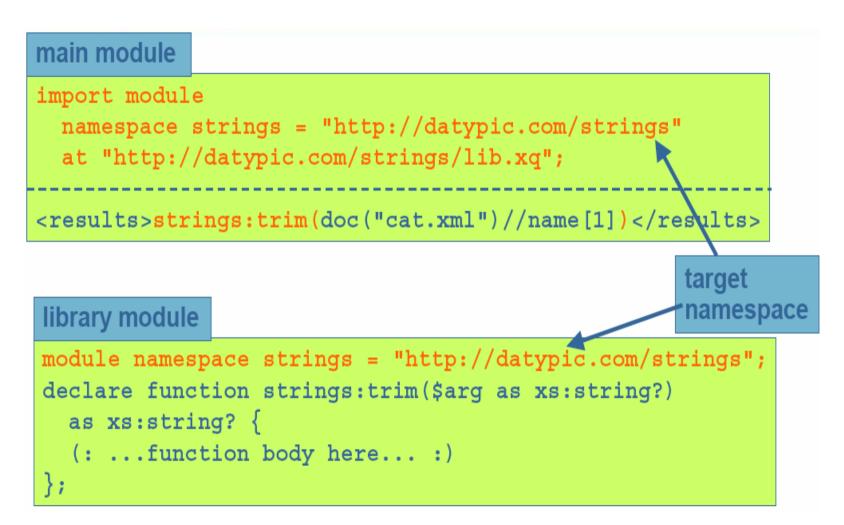
<description >{\$i/description}</description>

</item>



- Separate XQuery documents that contain function definitions.
- Why?
 - Reusing functions among many queries.
 - Defining standard libraries that can be distributed to a variety of query users.
 - Organizing and reducing the size of query modules.







- Declared and bound in the query prolog and used through the query.
- Can be
 - Referenced in a function that is declared in that module.
 - Referenced in other modules that import the module.

```
declare variable $maxNumItems := 3;
declare variable $ordDoc := doc("ord.xml");
for $item in
   $ordDoc//item[position() <= $maxNumItems]
return $item
```

Advantages of XQuery



- It is easy to learn if knowledge of SQL and XPath is present.
- When queries are written in XQuery, they require less code than queries written in XSLT do.
- XQuery can be used as a strongly typed language when the XML data is typed, which can improve the performance of the query by avoiding implicit type casts and provide type assurances that can be used when performing query optimization.
- XQuery can be used as a weakly typed language for untyped data to provide high usability.
- Because XQuery requires less code to perform a query than does XSLT, maintenance costs are lower.
- XQuery is supported by major database vendors.

Capabilities of XQuery



- Selecting information based on specific criteria
- Filtering out unwanted information
- Searching for information within a document or set of documents
- Joining data from multiple documents or collections of documents
- Sorting, grouping, and aggregating data
- Transforming and restructuring XML data into another XML vocabulary or structure
- Performing arithmetic calculations on numbers and dates
- Manipulating strings to reformat text

Uses of XQuery



- Extracting information from a relational database for use in a web service
- Generating reports on data stored in a database for presentation on the Web as XHTML
- Searching textual documents in a native XML database and presenting the results
- Pulling data from databases or packaged software and transforming it for application integration
- Combining content from traditionally non-XML sources to implement content management and delivery
- Ad hoc querying of standalone XML documents for the purposes of testing or research

Tutorial Overview

Session 1

XQuery Overview – Sherif SPARQL Overview – Axel XSPARQL: a combined language – Axel

Session 2 XQuery implementations – Sherif SPARQL implementations – Sherif XSPARQL implementations – Axel (optional) Compression formats for XML+RDF: EXI+HDT – Sherif

Q/A - Discussion





XQuery Implementations



Australian Government **Department of Broadband, Communications** and the Digital Economy

Australian Research Council

NICTA Funding and Supporting Members and Partners

Queensland Government









OF QUEENSLAND

THE UNIVERSITY OF SYDNEY

XQuery Processors



- Native XML/XQuery Processors: The implementations of this approach make use of storage models, indexing and querying mechanisms that have been designed specifically for XML data.
- Streaming XQuery Processors: The implementations of this approach receive the XML data in the form of continuous streams of tokens and apply on the fly the query processing functionalities over them.
- Relational XQuery Processors: The implementations of this approach makes use of the relational indexing and querying mechanisms for querying the source XML data..





- A native XML database which is able to store and query XML documents.
- It uses the tree-based query algebra (TAX) which considers collections of ordered labelled trees as the basic unit of manipulation
- Each operator takes one or more sets of trees as input and produces a set of trees as output.
- The query plan tree is evaluated by pipelining one operator at a time.
- The query execution heavily depends on structural joins.



- Another native XML database which clusters subtrees of XML documents into physical records of limited size.
- The XML data tree is partitioned into small subtrees and each subtree is stored into a data page.
- The query execution engine consists of an iterator-based implementation of algebraic operators which process ordered sequences of tuples.

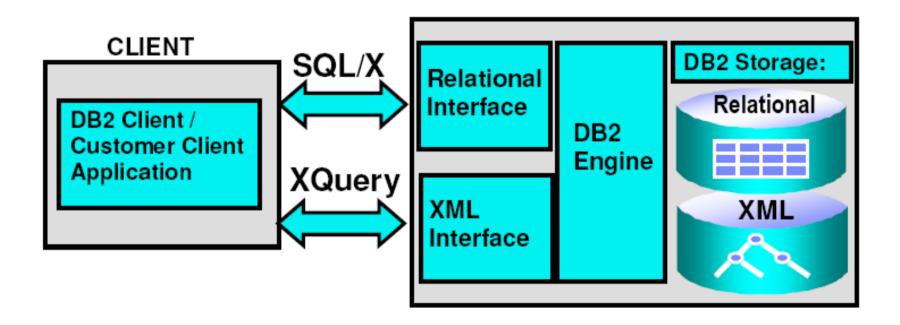




- A native XML support and XQuery implementation in IBM DB2.
- It uses a new XML data type which can be used like any other SQL type.
- A column of type XML can hold one well-formed XML document for every row of the table.
- Relational and XML data are stored differently while the relational columns are stored in traditional row structures, the XML data is stored in hierarchical structures.
- It uses Structural Indexes, Value Indexes and Full-text Indexes.
- Supports interfaces for both SQL/XML and XQuery.

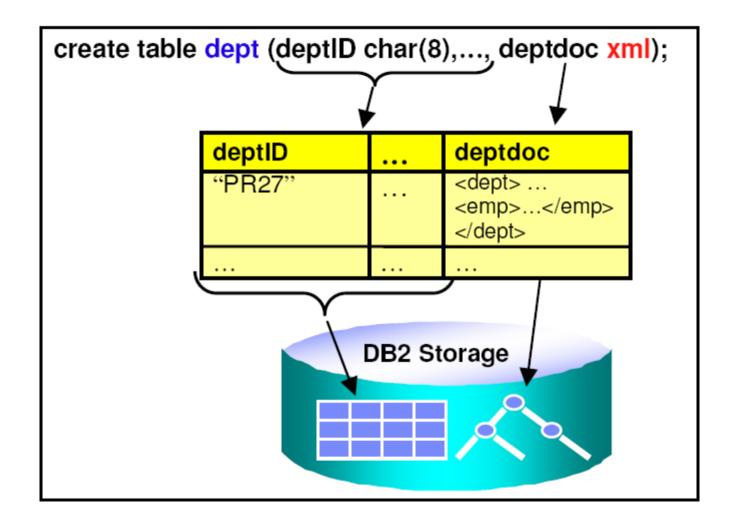






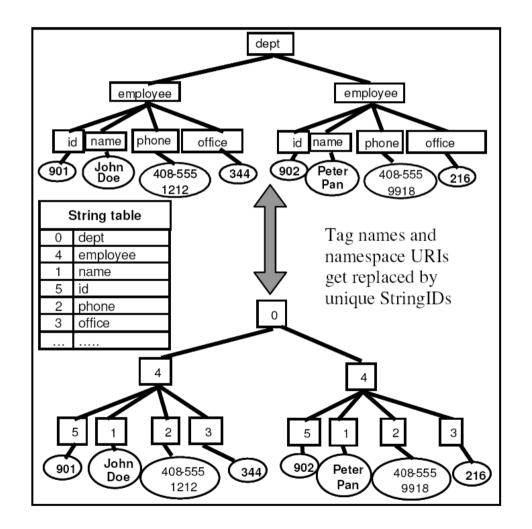




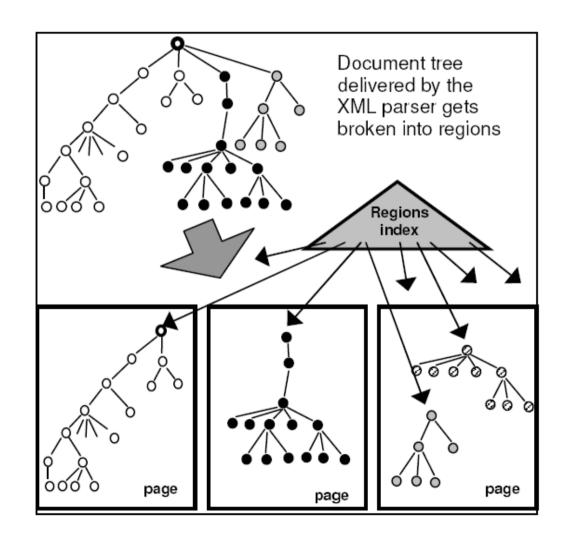


DB2/System RX







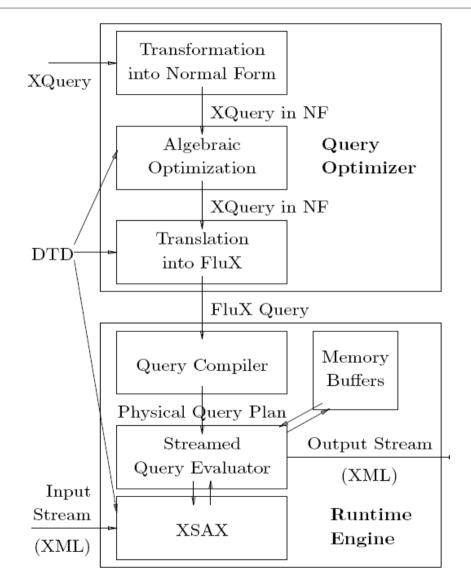




- A streaming XQuery processor that is based on an internal query language called FluX which extends the main structures of XQuery by introducing a construct for event-based query processing.
- The buffer size is then optimized by analyzing the schema constraints of the XML document as well as the query syntax.
- FluX queries are transformed into physical query plans which are translated into executable JAVA code or interpreted and executed using the Streamed Query Evaluator.

FluXQuery





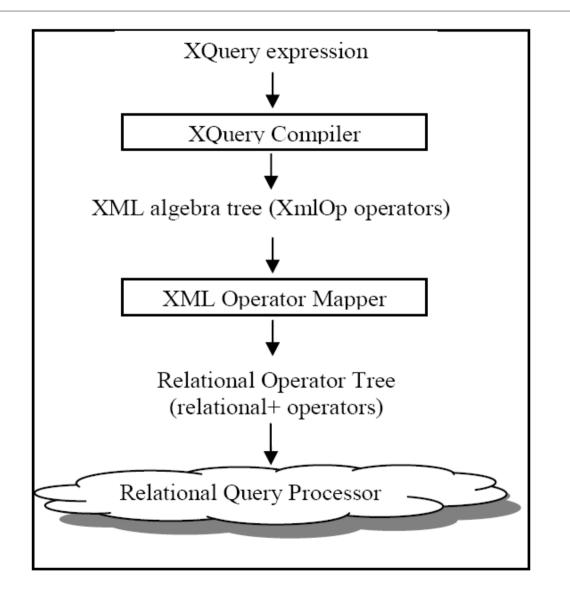


- Introduces native storage for XML data as a new, rich data type called XML. A table may contain one or more columns of type XML wherein both rooted XML trees and XML fragments can be stored.
- The hierarchical nature of the XML data is modelled as parent-child relationship using a node labelling scheme.
- XQuery expressions are compiled into a query trees that can be optimized and executed by the relational query processor.
- It uses a set of relational operators which is extended with additional operators for XML processing named as relational+ operators.
- The XML operator mapper recursively traverses the XML algebra tree and for each operator a relational operator sub-tree is generated and inserted into the overall relational operator tree for the XQuery expression.

NICTA Copyright 2012 Shankar Pal et al. XQuery Implementation in a Relational Database System. VLDB, 2005.

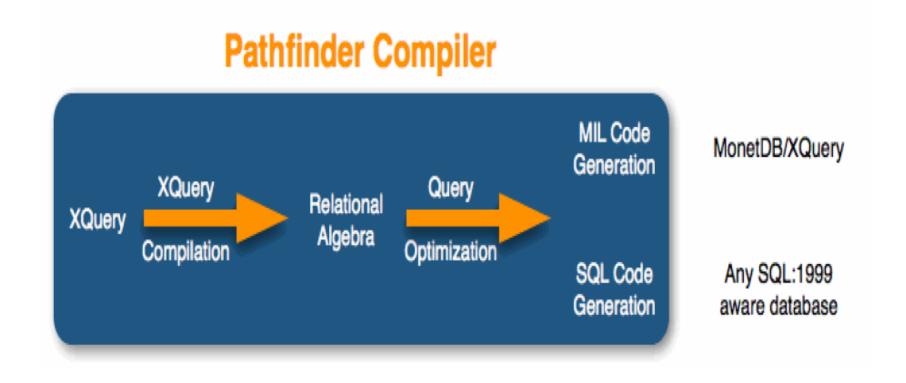
SQL Server





NICTA

A purely relational XQuery processor

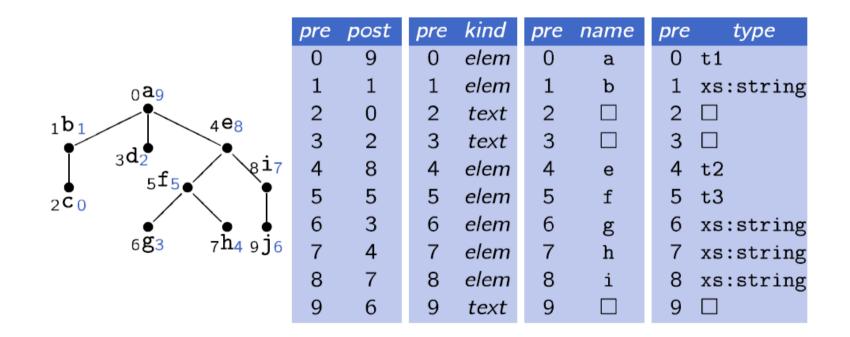


ICTA Copyright 2012

From imagination to impact



Relational XML Encoding





Relational Algebra

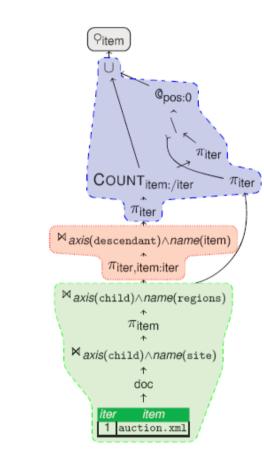
- π column projection, renaming
- σ row selection
- $\dot{\cup}$, \ disjoint union, difference
- δ duplicate elimination
- 🛛 equi-join
- × Cartesian product
- ρ row numbering
- 리 staircase join
- ε , au element/text node construction
- arithmetic/comparison/Boolean operator *
- No tree pattern matching or similar operators involved here
- This algebra is efficiently implementable on (top of) SQL hosts



Relational XQuery Compliaition

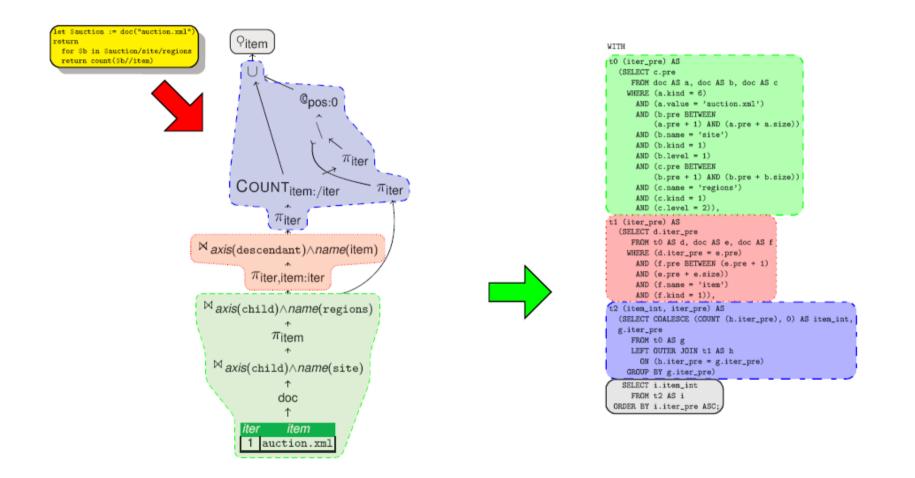
let \$auction := doc("auction.xml")
return
for \$h in \$custion/site/mensions

for \$b in \$auction/site/regions
return count(\$b//item)



NICTA Copyright Torsten Grust, Sherif Sakr, Jens Teubner: XQuery on SQL Hosts. VLDB, 2004





IICTA Copyright 2012 Torsten Grust, Sherif Sakr, Jens Teubner: XQuery on SQL Hosts. VLDB, 2004

Other Projects



- Galax
 - <u>http://www.galaxquery.org/</u>
- MXQuery
 - <u>http://mxquery.org/</u>
- Zorba
 - <u>http://www.zorba-xquery.com/html/index</u>
- BaseX
 - http://basex.org/
- Xyleme
 - <u>http://www.xyleme.com/</u>
- eXist
 - <u>http://exist.sourceforge.net/</u>
- Saxon
 - <u>http://saxon.sourceforge.net/</u>
- MarkLogic
 - http://developer.marklogic.com/

XML Database Benchmarks



- XBench
 - <u>http://se.uwaterloo.ca/~ddbms/projects/xbench/index.html</u>
- XMach
 - <u>http://dbs.uni-leipzig.de/en/projekte/XML/XmlBenchmarking.html</u>
- XMark
 - <u>http://www.xml-benchmark.org/</u>
- XOO7
 - http://www.comp.nus.edu.sg/~ebh/XOO7/main.html
- TPoX
 - <u>http://tpox.sourceforge.net/</u>
- XPathMark
 - <u>http://sole.dimi.uniud.it/~massimo.franceschet/xpathmark/</u>
- MemBeR
 - <u>http://ilps.science.uva.nl/Resources/MemBeR/member-generator.html</u>
- XSelMark
 - <u>http://xselmark.sourceforge.net/</u>

Tutorial Overview

Session 1



XQuery Overview – Sherif SPARQL Overview – Axel XSPARQL: a combined language – Axel Compression formats for XML+RDF: EXI+HDT – Sherif

Session 2

XQuery implementations - Sherif

SPARQL implementations – Sherif

XSPARQL implementations – Axel

(optional) Compression formats for XML+RDF: EXI+HDT – Sherif

Q/A - Discussion





Australian Government **Department of Broadband, Communications** and the Digital Economy

Australian Research Council

NICTA Funding and Supporting Members and Partners

Queensland Government









THE UNIVERSITY OF SYDNEY

Two Alternatives to Store RDF Data



- Relational RDF Stores
 - Takes advantage of 30+ years of R&D
 - A schema design exercise?
 - Requires a SPARQL-to-SQL layer
 - Relative flexibility in choosing the actual back-end
- Native RDF Stores
 - Takes advantage of 30+ years of R&D
 - Highly tuned for RDF data

Relational RDF Stores



• Vertical (triple) table stores: where each RDF triple is stored

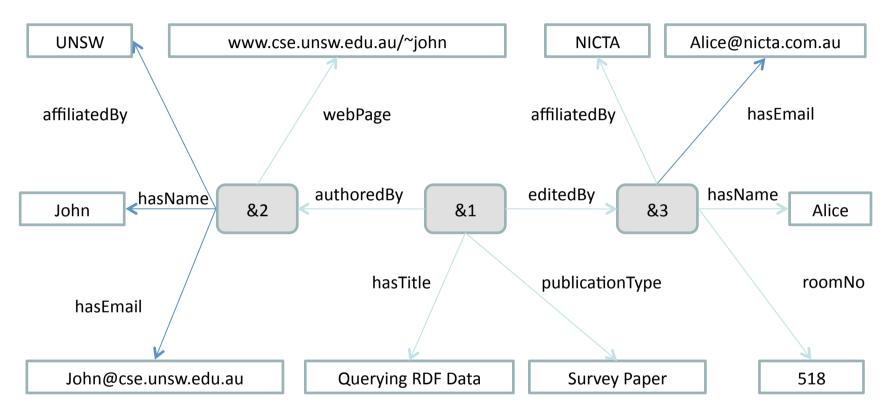
directly in a *three-colum*n table (subject, predicate, object).

- Property (n-ary) table stores: where multiple RDF properties are modeled as *n-ary* table columns for the same subject.
- Horizontal (binary) table stores: where RDF triples are modeled as one horizontal table or into a set of vertically partitioned binary

tables (one table for each RDF property).

Relational RDF Stores





SELECT ?ZWHERE { ?X hasTitle "Querying RDF Data". ?X publicationType "Survey Paper".
?X authoredBy ?Y.?Y webPage ?Z. }



Subject	Predicate	Object
ld1	publicationType	Survey Paper
ld1	hasTitle	Querying RDF Data
ld1	authoredBy	ld2
ld2	hasName	John
ld2	affiliatedBy	UNSW
ld2	hasEmail	John@cse.unsw.edu.au
ld2	webPage	www.cse.unsw.edu.au/~john
ld1	editedBy	ld3
ld3	hasName	Alice
ld3	affiliatedBy	NICTA
ld3	hasEmail	Alice@nicta.com.au
ld3	roomNo	518

Select T3.Object From Triples as T1, Triples as T2, Triples as T3, Triples as T4 Where T1.Predicate="publicationType" and T1.Object="Survey Paper" and T2.predicate="hasTitle" and T2.Object="Querying RDF Data" and T3.Predicate="webPage" and T1.subject=T2.subject and T4.subject=T1.subject and T4.Predicate="authoredBy" and T4.Object = T3.Subject

No schema changes as new triples are added to the store SPARQL queries result in equivalent SQL queries involving multiple self-joins

Property Table Stores



Publication

ID	publicationType	hasTitle	authoredBy	editedBy
ld1	Survey Paper	Querying RDF Data	ld2	id3

Person

ID	hasName	affiliatedBy	hasEmail	webPage	roomNo
ld2	John	UNSW	John@cse.unsw.edu.au	www.cse.unsw.edu.au/~john	
ld3	Alice	NICTA	Alice@nicta.com.au		518

Select Person.webPage
From Person, Publication
Where Publication.publicationType = "Survey Paper"
and Publication.hasTitle = "Querying RDF Data"
and Publication.authoredBy = Person.ID

Implicit or explicit knowledge is required to build the property tables

Horizontal (Binary) Table Stores



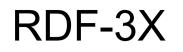
publicationType hasTit		le			webPa	age			
ld1	Survey Paper	ld1	Querying RDF Data		ld2	www.cse.un	isw.edu.au/∼j	.edu.au/~john	
hasNa	me		affiliate	dBy		autho	redBy	editedE	Зу
ld2	John		ld2	UNS	N	ld1	ld2	ld1	ld3
ld3	Alice		ld3	NICT	A				·
hasEm	nail			roomN	0				
ld2	John@cse.unsw	v.edu.au	ld3 518						
ld3	Alice@nicta.con	n.au							
			From Where and ha and pu and pu	authore publica sTitle.va ublicatio ublicatio	onType, I dBy, web tionType lue = "Qu nType.ID nType.ID	Page	e.ID redBy.ID	.11	

Does this work for Dbpedia which has approx. 65K properties?

RDF-3X



- An RDF query engine which tries to overcome the criticism that triples stores incurs too many expensive self-joins by creating the exhaustive set of indexes and relying on fast processing of merge joins.
- The physical design of RDF-3x is workload-independent and eliminates the need for physical-design tuning by building indexes over all 6 permutations of the three dimensions that constitute an RDF triple.
- Additionally, indexes over count-aggregated variants for all three two-dimensional and all three one-dimensional projections are created.





- The query processor uses the full set of indexes on the triple tables to rely mostly on *merge joins* over sorted index lists.
- The query optimizer relies upon its cost model in finding the lowest-cost execution plan and mostly focuses on join order and the generation of execution plans.
- Additionally, indexes over count-aggregated variants for all three twodimensional and all three one-dimensional projections are created.
- It relies on two kinds of statistics:
 - Specialized histograms which are generic and can handle any kind of triple patterns and joins. The disadvantage of histograms is that it assumes independence between predicates.
 - *Frequent join paths* in the data which give more accurate estimation.

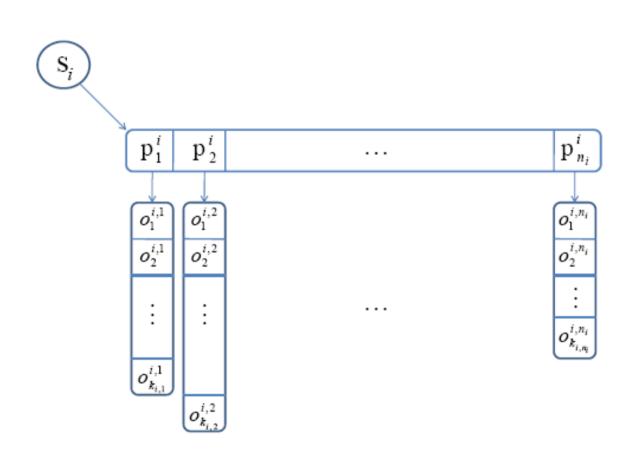


- RDF storage scheme with main focuses on scalability and generality in its data storage, processing and representation.
- Hexastore is based on the idea of indexing the RDF data in a multiple indexing scheme.
- Each RDF element type have its special index structures built around it.
- Every possible ordering of the importance or precedence of the three elements in an indexing scheme is materialized.
- In total, *six* distinct indices are used for indexing the RDF data. These indices materialize all possible orders of precedence of the three RDF elements.
- A clear disadvantage of this approach is that it features a worst-case *five-fold* storage increase in comparison to a conventional triples table.

Cathrin Weiss, Panagiotis Karras, and Abraham Bernstein: Hexastore: sextuple indexing for semantic web NICTA Copyright 2012 data management. PVLDB 1(1), 2008

Hexastore





An Example SPO Index of Hexastore

From imagination to impact

Oracle RDF



- Oracle introduced an Oracle-based SQL table function *RDFMATCH* to query RDF data.
- The results of *RDFMATCH* table function can be further processed by SQL's rich querying capabilities and seamlessly combined with queries on traditional relational data.
- The core implementation of *RDFMATCH* query translates to a selfjoin query on triple-based RDF table store.
- The resulting query is executed efficiently by making use of B-tree indexes as well as creating *materialized join views* for specialized *subject-property*.



- Subject-Property Matrix materialized join views are used to minimize the query processing overheads that are inherent in the canonical triple-based representation of RDF.
- The materialized join views are *incrementally* maintained based on *user demand* and *query workloads*.
- A special module is provided to analyze the table of RDF triples and estimate the size of various materialized views, based on which a user can define a subset of materialized views.



```
SELECT t.r reviewer, t.c conf, t.a age
FROM TABLE(RDF_MATCH(
    '(?r rdf:type Student)
    (?r ReviewerOf ?c)
    (?r Age ?a)',
    RDFModels ('reviewers'),
    NULL, NULL)) t
WHERE t.a < 25;</pre>
```

Workload Independent Property Tables



- The approach provides a *tailored* schema for each RDF data based on two main parameters:
 - The *Support threshold* which represents a value to measure the strength of correlation between properties in the RDF data.
 - The *null threshold* which represents the percentage of null storage tolerated for each table in the schema.
- The approach involves two phases
 - The *clustering* phase scans the RDF data to automatically discover groups of related properties (i.e., properties that always exist together for a large number of subjects).
 - The *partitioning* phase goes over the formed clusters and balances the trade-off between storing as many RDF properties in clusters as possible while keeping null storage to a minimum based on the null threshold.

NCJusting J.2Levandoski and Mohamed En Mokbel: RDF Data-Centric Storage. ICWS, 2009



- The main focus of this approach is to improve the performance for *path queries* by extracting all reachable path expressions for each resource and store them.
- There is no need to perform join operations unlike the flat tripe stores or the property tables approach.
- The RDF graph is divided into subgraphs and then each subgraph is stored by applicable techniques into distinct relational tables. More precisely, all classes and properties are extracted from RDF schema data, and all resources are also extracted from RDF data.
- Each extracted item is assigned an identifier and a path expression and stored in corresponding relational table.

Akiyoshi Matono, Toshiyuki Amagasa, Masatoshi Yoshikawa, and Shunsuke Uemura: A Path-based NICTA Copyright 2012 Relational RDF Database. ADC, 2005



- The triples table is rewritten into *n* two-column tables where *n* is the number of unique properties in the data.
- In each binary table, the first column contains the subjects that define that property and the second column contains the object values for those subjects while the subjects that do not define a particular property are simply omitted from the table for that property.
- Each table is sorted by subject, so that particular subjects can be located quickly, and that fast merge joins can be used to reconstruct information about multiple properties for subsets of subjects.

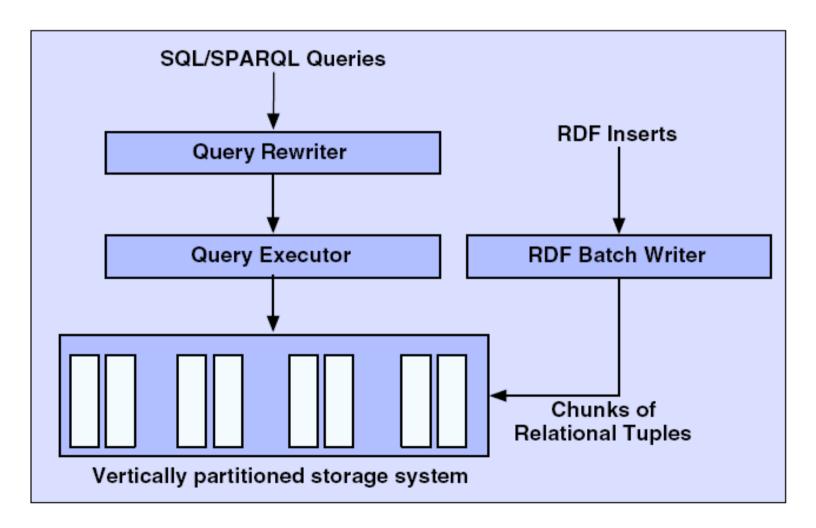
Daniel J. Abadi, Adam Marcus, Samuel Madden, and Kate Hollenbach: SW-Store: a vertically partitioned DBMS for Semantic Web data management. VLDB J. 18(2), 2009



- For a *multi-valued* attribute, each distinct value is listed in a successive row in the table for that property.
- The implementation of SW-Store relies on a column-oriented DBMS,
 C-store, to store tables as collections of columns rather than as collections of rows.
- The advantage of this approach is that while property tables need to be carefully constructed so that they are wide enough but not too wide to independently answer queries, the algorithm for creating tables in the vertically partitioned approach is straightforward and need not change over time.
- The main disadvantages of this approach are: the increased cost of inserts new triples and the cost of tuple reconstruction

NICTA Copyright 2012



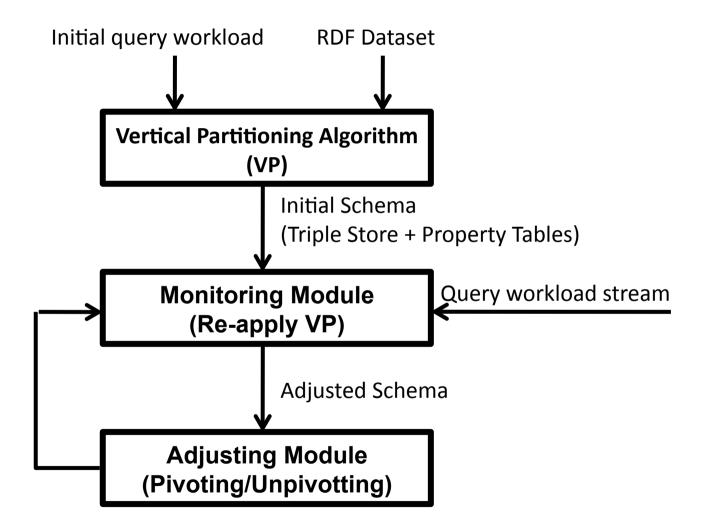




- A two-phase approach for designing efficient tailored but flexible storage solution for RDF data based on its query workload.
- A workload-aware vertical partitioning phase.
 - Reduces the number of join operations in the query evaluation process
- An *automated adjustment* phase
 - Maintains the efficiency of the performance of the query processing by adapting the underlying schema to cope with the dynamic nature of the query workloads
 - Rows pivoting and uninviting

AdaptRDF







(Upper				Lower		er)		
		a_6	a_2	a_3	a_7	a_{10}	a_8	a_1	a_9	a_4	a_5	
	a_6	90	30	25	25	15	15	10	0	0	0	
	a_2	30	105	60	40	20	15	15	0	0	0	
	a_3	25	60	40	30	15	15	10	0	0	0	
J	a_7	25	40	30	65	25	15	10	0	0	0	l
١	a_{10}	15	20	15	25	70	10	10	0	0	0	Ì
	a_8	15	15	15	15	10	60	10	18	18	18	
	a_1	10	15	10	10	10	10	50	18	18	18	
	a_9	0	0	0	0	0	18	18	40	40	40	
	a_4	0	0	0	0	0	18	18	40	40	40	
l	a_5	0	0	0	0	0	18	18	40	40	40	J





Subject	Predicate	Object		
ld1	publicationType	Survey Paper		
ld1	hasTitle	Querying RDF Data		
ld1	authoredBy	ld2		
ld2	hasEmail	John@cse.unsw.edu.au		
ld2	webPage	www.cse.unsw.edu.au/~john		
ld1	editedBy	ld3		
ld3	hasEmail	Alice@nicta.com.au		

Subject	hasName	affiliatedBy	roomNo
ld2	John	UNSW	
ld3	Alice	NICTA	518

(a) Pivot (hasName, affiliatedBy, roomNo)

Subject	Predicate	Object			
ld1	publicationType	Survey Paper			
ld1	hasTitle	Querying RDF Data			
ld1	authoredBy	ld2			
Id2	hasEmail	John@cse.unsw.edu.au			
Id2	webPage	www.cse.unsw.edu.au/~john			
ld1	editedBy	ld3			
Id3	hasEmail	Alice@nicta.com.au			
Id3	roomNo	518			

Subject	hasName	affiliatedBy
Id2	John	UNSW
ld3	Alice	NICTA

(b) Unpivot (roomNo)

RDF Benchmarks



- Lehigh University Benchmark (LUBM)
 - <u>http://swat.cse.lehigh.edu/projects/lubm/</u>
 - (University domain)
- Berlin SPARQL Benchmark (BSBM)
 - <u>http://www4.wiwiss.fu-berlin.de/bizer/berlinsparqlbenchmark/</u>
 - E-Commerce domain
- The SP²Bench SPARQL Performance Benchmark
 - <u>http://dbis.informatik.uni-freiburg.de/index.php?project=SP2B</u>
 - DBLP Scenario
- The DBpedia SPARQL Benchmark
 - <u>http://aksw.org/Projects/DBPSB</u>
 - Wikipedia + YAGO





- The benchmark implements two relational RDF storage solutions triple-store and vertically-partitioned – in a MonetDB/SQL, a fullyfunctional open source column-store, and a well-known – for its performance – commercial row-store DBMS.
- The vertically-partitioned approach does not outperform triple-store when both are implemented in a row-store engine.
- The vertically-partitioned approach outperforms triple-store when both are implemented in a column-store. The processing efficiency of column-stores is particularly suited for RDF data management applications.

Benchmarkings (1)



- Once the proper clustered indices are used in a row-store engine, the triple-store performs better than the vertically-partitioned approach.
- There is a potential of a scalability problems for the verticallypartitioned approach when the number of properties in an RDF dataset is high. With a larger number of properties, the triple-store solution manages to outperform the vertically-partitioned approach also on a column-store engine.





- The benchmark evaluates the approaches of relational processing for RDF Queries: *triple stores* (*TS*), *binary table stores* (*BS*), *property table stores* (*PS*) and *traditional relational stores* (*RS*).
- The experimental evaluation is done on top of the SP²Bench benchmark and compares the following metrics: *loading time*, *storage costs and query performance*.
- The *RS* scheme is the fastest due to the less required number of insert tuple operations. Similarly, the *TS* requires less loading time than *BS* since the number of inserted tuples and updated tables are smaller for each triple.



- The *RS* scheme represents the cheapest approach because of the normalized design and the absence of any data redundancy. The *BS* scheme represents the most expensive approach due to the redundancy of the ID attributes for each binary table.
- There is no clear winner between the triple store (TS) and the binary table (BS) encoding schemes. Triple store (TS) with its simple storage and the huge number of tuples in the encoding relation is still very competitive to the binary tables encoding scheme because of the full set of B-tree physical indexes over the permutations of the three encoding fields (subject; predicate; object).

Hooran MahmoudiNasab and Sherif Sakr: An Experimental Evaluation of Relational RDF Storage and From imagination to impact Querying Techniques. BenchmarX, 2010



- The query performance of the (BS) encoding scheme is affected badly by the increase of the number of the predicates In the input query. It is also affected by the subject-object or object-object type of joins where no index information is available for utilization. Such problem could be solved by building materialized views over the columns of the most frequently referenced pairs of attributes.
- Although their generality, there is still a clear gap between the query performance of the (TS) and (BS) encoding schemes in comparison with the tailored relational encoding scheme (RS) of the RDF data.

Matrix "Bit" loaded

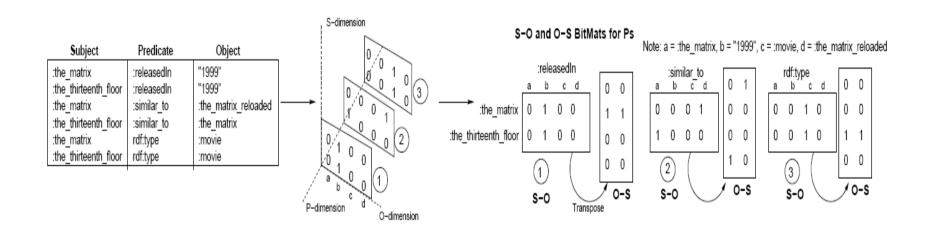


- A compressed *bit-matrix* structure for storing huge RDF graphs.
- The RDF data is represented by a 3D bit-cube, where each dimension of the *bitcube* represents subjects, predicates, and objects.
- The volume of this *bitcube* is Vs × Vp × Vo.
- Each cell in the *bitcube* represents a unique RDF triple that can be formed by the combination of S, P, O positions which are the coordinates of that bit.
- For each matrix, it stores two *bitarrays* row and column bitarray which give a condensed representation of all the non-empty row and column values in the given BitMat.
- It employs an initial *pruning* technique, followed by a *variable-binding-matching* algorithm on *BitMats* to produce the final results.
- It does not build intermediate join tables and works directly on the compressed data.

Medha Atre, Vineet Chaoii, Mohammed J. Zaki, James A. Hendler: Matrix "Bit" loaded: a scalable From imagination to impact lightweight join query processor for RDF data . WWW, 2010

Matrix "Bit" loaded







- Developed by the Hewlett Packard Semantic Web Lab.
- Jena is a Semantic Web application framework for Java.
- It provides a programming environment for semantic technologies such as RDF(S) and OWL that supports SPARQL queries and a rule-based inference engine.
- Its engine is a combination of in-memory and native.
- A native, on-disk implementation of Jena is the Jena TDB triple store.



- Triples are stored in the form of a *dictionary*. Every node is given a unique ID and instead of storing the whole node, only their ID is put in the triple store.
- Jena supports different RDF syntaxes: RDF/XML, N3, N-Triples, Turtle.
- Jena SDB version can use PostgreSQL, MySQL, Oracle or MS SQL Serveras its backend.



- AllegroGraph is a persistent graph database developed by Franz Inc. that uses in-memory utilization in combination with native engine in order to maintain billions of quads.
- AllegroGraph supports named graphs. It has an optimised indexing system consisting of different combinations of S, P, O, G and I.
- It has a mechanism of automatic index optimisation that optimises indices based on how much they are used.
- Querying can be done through SPARQL 1.1 queries, with support for full-text, geospatial and geotemporal search.
- It is also compatible with a lot of other Franz technologies such as *RacerPro*, *Pepito*, *TopBraid Composer*, etc. to improve performance and usability.



- In June 2011, AllegroGraph announced at Semtech conference a load and query of 310 Billion triples as part of a joint project with Intel.
- In August 2011, AllegroGraph announced that with the help of Stillwater SC and Intel, they achieved the industry's first load and query of 1 Trillion RDF Triples. Total load was 1,009,690,381,946 triples in just over 338 hours for an average rate of 829,556 triples per second.
- It is in late-stage development on a clustered version that will push storage into trillions of triples.





- Virtuoso is an object-relational SQL database that supports SPARQL embedded into SQL for querying RDF data in the Virtuoso database.
- Querying can be done by using SPARQL queries. There is however no SPARQL endpoint, only an SQL command line interface through which you can send SPARQL 1.1 queries to the Virtuoso repository.
- There is also a possibility to send the queries using ODBC, JDBC, ADO.NET or OLE DB.
- Full-text search and geospatial search are also supported.
- The reasoner that is implemented in the Virtuoso store is a backward chaining OWL reasoner, but reasoning can be added through the Jena framework.





- Mulgara was formerly known as Kowari.
- An Open Source, massively scalable, transaction-safe, purpose-built database for the storage, retrieval and analysis of metadata.
- It supports RDF and OWL.
- It is optimised for large amounts of triples, for metadata in particular, but it keeps low memory requirements.
- Triples in the Mulgara store are stored natively (on-disk), both locally or distributed.
- Supports SPARQL in addition to a language called *TQL* where Full-text search functionality is also available.
- Mulgara supports: RDF/XML, N-Triples, Turtle.
- Mulgara has a reasoner that supports: DAML+OIL, RDF(S) and OWL

Other Projects

- Sesame
 - <u>http://www.openrdf.org/</u>
- 3Store
 - <u>http://www.aktors.org/technologies/3store/</u>
- 4store
 - http://4store.org/
- BigOWLIM
 - <u>http://www.ontotext.com/owlim/editions</u>
- Bigdata
 - http://www.systap.com/bigdata.htm
- BrightstarDB
 - <u>http://www.brightstardb.com/</u>
- Apache Fuseki
 - <u>http://incubator.apache.org/jena/documentation/serving_data/</u>



Other Projects

- Parliament
 - <u>http://parliament.semwebcentral.org/</u>
- OpenAnzo
 - http://www.openanzo.org/
- StrixDB
 - <u>http://www.strixdb.com/</u>
- Redstore
 - <u>http://www.aelius.com/njh/redstore/</u>
- Meronymy
 - <u>http://www.meronymy.com/</u>
- Neo4j
 - <u>http://neo4j.org/</u>
- rdfstore-js
 - https://github.com/antoniogarrote/rdfstore-js







- The benchmark compared between: *Jena TDB*, *AllegroGraph*, *OWLIM-SE*, *Sesame* and *Mulgara*.
- The experiments have been conducting using the LUMB, BSBM and SP²Bench benchmarks.
- The comparison metrics are: Loading time, Query time and ease of use.
- The experiments reported that the *Mulgara* was the easies to install, run and use.
- OWLIM has the best performance for loading time and query time followed by *Mulgara* then *Jena*.
- *AllegroGraph* is the most robust tested system.



- A horizontally scalable RDF database system.
- It installs a best-of-breed RDF-store on a cluster of machines (they use RDF-3X since they found this to be the fastest single-node RDF-store in their benchmarking).
- It partitions an RDF data set across the data stores. Instead of randomly assigning triples to partitions using hash partitioning, they take advantage of the fact that RDF uses a graph data model, they use an optimized graph partitioning algorithm.
- Triples that are close to each other in the RDF graph to be stored on the same machine and a smaller amount of network communication at query time is required.



- In order to maximize the percentage of query processing that can be done in parallel, they allow some overlap of data across partitions.
- They use a method for automatic decomposition mechanism of queries into chunks that can be performed independently, with zero communication across partitions. These chunks are then reconstructed using the Hadoop MapReduce framework (*HadoopDB*).
- The performance is up to three orders of magnitude more efficient than popular multi-node RDF data management systems.

Azza Abouzeid, Kamil Bajda-Pawlikowski, Daniel J. Abadi, Alexander Rasin and Avi Silberschatz: *HadoopDB: An* NICTA Copyrig^{ID} Architectural Hybrid of MapReduce and DBMS Technologies for Analytical Workloads. PVLDB 2(1), 2009

Distributed Processing of RDF/SPARQL

•The master node also serves as the interface for SPARQL queries. It accepts queries and analyzes them closely.

• If it determines that the SPARQL pattern can be searched for completely in parallel by each worker in the cluster, then it sends the pattern to each node in the cluster for processing.

•If, it determines that the pattern requires some coordination across workers during the search, it decomposes the query into subgraphs that can be searched for in isolation, and ships these subgraphs to the worker nodes.

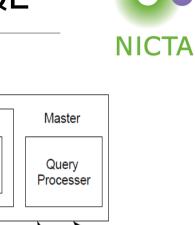
Data Partitioner Triple Graph Querv Partitioner Placer Processer Worker 1 Worker 2 RDF-store RDF-store RDF-store

Data

replicator

Data

replicator



Hadoop

Worker n

Data

replicator

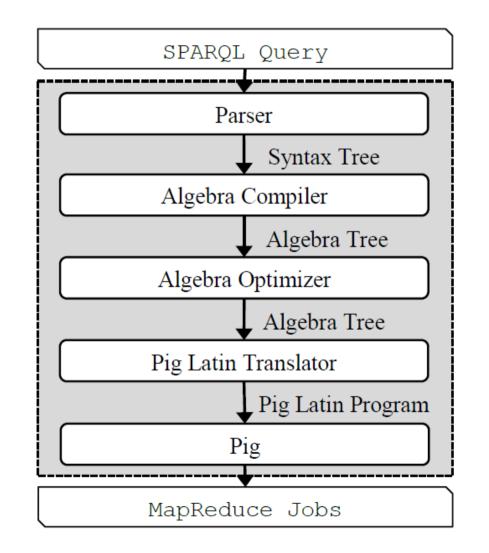




- A system which gives us the opportunity to process complex SPARQL queries on a MapReduce cluster.
- SPARQL queries are translated into *Pig Latin*, a data analysis language developed by Yahoo! Research.
- Pig Latin programs are executed by a series of MapReduce jobs on a Hadoop cluster.
- PigSPARQL offers not only a declarative way of specifying the transformation part, but also a scalable implementation of the whole ETL-process on a MapReduce cluster.

PigSPARQL





Alexander Schätzl, Martin Przyjaciel-Zablock and Georg Lausen: PigSPARQL: Mapping SPARQL to Pig From imagination to impact Latin. SWIM, 2011

HBase vs MySQL Cluster



- A comparison between *HBase* and *MySQL Cluster* for SPARQL query processing.
- An experimental comparison of the two proposed approaches on a cluster of commodity machines LUMB.
- Both approaches were up to the task of efficiently storing and querying large RDF datasets.
- The the HBase solution was capable of dealing with larger RDF datasets and showed superior query performance and scalability.
- Cloud computing has a great potential for scalable Semantic Web data management.

Tutorial Overview

Session 1



XQuery Overview – Sherif SPARQL Overview – Axel XSPARQL: a combined language – Axel Compression formats for XML+RDF: EXI+HDT – Sherif

Session 2 XQuery implementations – Sherif SPARQL implementations – Sherif XSPARQL implementations – Axel (optional) Compression formats for XML+RDF: EXI+HDT – Sherif

Q/A - Discussion



XML Compression



 Australian Government
 Department of Broadband, Communications and the Digital Economy

Australian Research Council

NICTA Funding and Supporting Members and Partners

Queensland Government





QUT

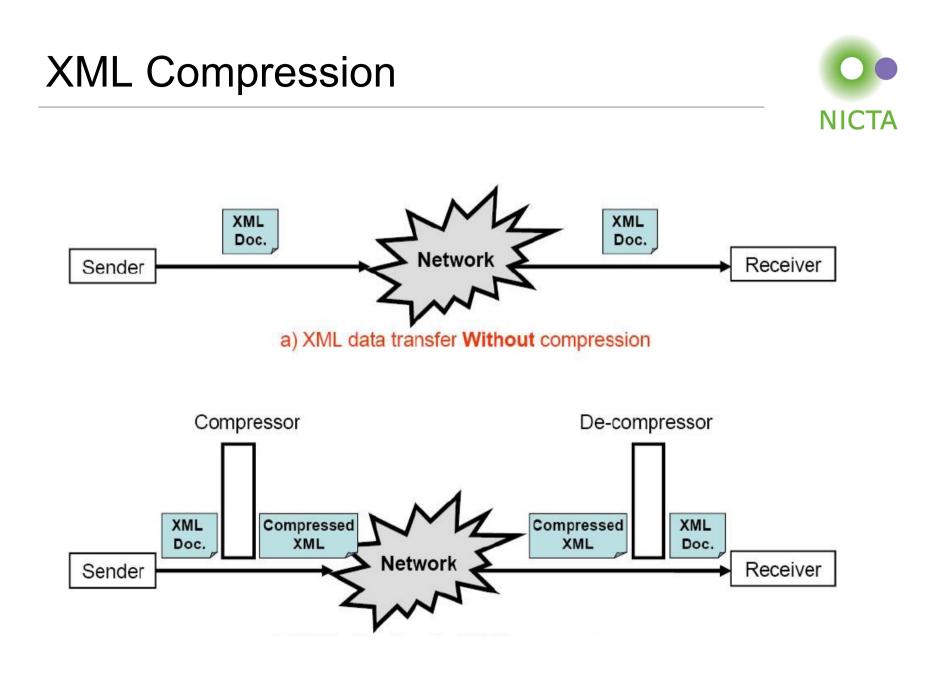


THE UNIVERSITY OF QUEENSLAND

NICTA Copyright 2012

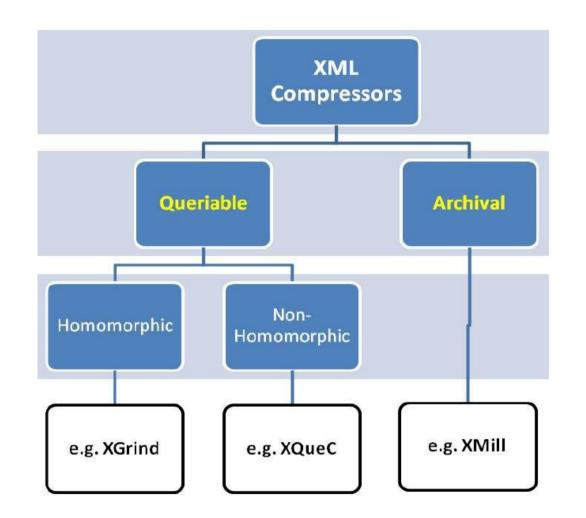
From imagination to impact

THE UNIVERSITY OF SYDNEY



XML Compressors





From imagination to impact



- General Text Compressors: Since XML data are stored as text files, these compressors use the traditional general purpose text compression tools. They are XML-Blind, i.e. they treat XML documents as usual plain text documents and thus apply the traditional text compression techniques.
- *XML Conscious Compressors:* This group of compressors are designed to take the advantage of the awareness of the XML document structure in order to achieve better compression ratios over the general text compressors.
 - Schema dependent compressors: where both of the encoder and decoder must have access to the document schema information to achieve the compression process.
 - Schema independent compressors: where the availability of the schema information is not required to achieve the encoding and decoding processes.



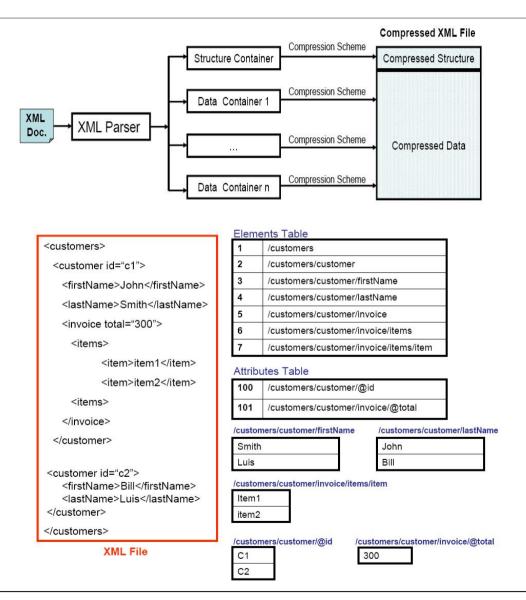
- **Non-Queriable (Archival) XML Compressors:** This group of the XML compressors does not allow any queries to be processed over the compressed format. The main focus of this group is to achieve the highest compression ratio. By default, general purpose text compressors belong to the non-queriable group of compressors.
- **Queriable XML Compressors:** This group of the XML compressors allow queries to be processed over their compressed formats. The compression ratio of this group is usually worse than that of the archival XML compressors. However, the main focus of this group is to avoid full document decompression during query execution.
 - Homomorphic compressors: where the original structure of the XML document is retained and the compressed format can be accessed and parsed in the same way of the original format.
 - Non-homomorphic compressors: where the encoding process of the XML document serrates the structural part from the data part. Therefore, the structure of the compressed format is different from the structure of the original XML document.



- Separates the structure from data and the grouping of the data values into homogenous containers based on their relative paths in the tree and their data types.
- Both of the structural and data value parts of the source XML document are collected and compressed separately.
- In the structure part, XML tags and attributes ar encoded in a dictionary-based fashion before passing it to a back-end general text compression scheme.
- In the data part, data values are grouped into homogenous and semantically related containers according to their path and data type.
 Each container is then compressed separately using specialized compressor that is ideal for the data type of this container.

XMill





VICTA Copyright 2012

http://www.cs.washington.edu/homes/suciu/XMLTK/xmill/www/





- XML-conscious compression scheme to support querying **without** the need for a full decompression of the compressed XML document.
- *XGrind* does not separate data from structure. It retains the original structure of the XML document.
- Element and attribute names are encoded using a dictionary-based encoding.
- Character data is compressed using semi-adaptive Huffman coding.
- The query processor of XGrind can only handle exact-match and prefixmatch queries on compressed values and partial-match and range queries on decompressed values.
- Several operations are not supported by *XGrind*, for example, non-equality selections in the compressed domain. Therefore, *XGrind* cannot perform any join, aggregation, nested queries, or construct operations.

NICTA Copyright 20 Pankaj M. Tolani, Jayant R. Haritsa, XGRIND: A query-friendly XML compressor, ICDE, 2002.

XML Compressors



- XMLPPM
 - <u>http://xmlppm.sourceforge.net/</u>
- XWRT
 - <u>http://sourceforge.net/projects/xwrt/</u>
- Exalt
 - <u>http://exalt.sourceforge.net/</u>
- Rngzip
 - <u>http://contrapunctus.net/league/haques/rngzip/</u>
- Benchmark
 - <u>http://xmlcompbench.sf.net/</u>

XML Fast Infoset



- <u>http://www.w3.org/XML/Binary/</u>
- An international standard that specifies a binary encoding format for XML documents.
- It aims to provide more efficient serialization than the text-based XML format.
- The underlying file format is with tag/length/value blocks.
- Text values of attributes and elements are therefore stored with length prefixes rather than end delimiters.
- An index table is built for most strings, which includes element and attribute names, and their values. This means that the text of repeated tags and values only appears once per document.

NICTA Copyright 2012 http://java.sun.com/developer/technicalArticles/xml/fastinfoset/





- W3C recommendation for a compact, high performance XML representation that is designed to work well for a broad range of applications.
- It aims of improving performance and significantly reducing bandwidth requirements without compromising efficient use of other resources such as battery life, code size, processing power, and memory.
- EXI uses a grammar-driven approach and a straightforward encoding algorithm and a small set of datatype representations. Therefore, EXI processors should be relatively simple and can be implemented on devices with limited capacity.



- EXI is schema *informed*, meaning that it can utilize available schema information to improve compactness and performance, but does not depend on accurate, complete or current schemas to work.
- A program module called an EXI processor is used by application programs to encode their structured data into EXI streams (EXI stream encoder) and/or to decode EXI streams to make the structured data accessible (EXI stream decoder).
- An EXI stream is an EXI Header followed by an EXI body. The EXI body carries the content of the document, while the EXI header communicates the options used for encoding the EXI body





- The building block of an *EXI body* is an *EXI event*. An EXI body consists of a sequence of EXI events representing an *EXI document* or an *EXI fragment*.
- The *EXI events* permitted at any given position in an EXI stream are determined by the *EXI grammar*. As is the case with XML, the events occur with nesting pairs of matching start element and end element events where any pair does not intersect with another except when it is fully contained in the other.
- The *EXI grammar* incorporates knowledge of the XML grammar and may be augmented and refined using schema information and fidelity options.





- Each EXI stream begins with an EXI header.
- The **EXI header** can identify EXI streams, distinguish EXI streams from text XML documents, identify the version of the EXI format being used, and specify the options used to process the body of the EXI stream.
- The EXI Options field within an EXI header is optional. Its presence is indicated by the value of the presence bit that follows distinguishing bits.
- When the *alignment* option is *byte-alignment*, padding bits of minimum length required to make the whole length of the header byte-aligned are added at the end of the header. On the other hand, there are no padding bits when the alignment in use is *bit-packed*.

[EXI Cookie] Distinguishing Bits for	Presence Bit E or EXI Options	EXI Format Version	[EXI Options]	[Padding Bits]
--------------------------------------	----------------------------------	-----------------------	---------------	----------------



- An EXI header MAY start with an **EXI Cookie** which is a four byte field that serves to indicate that the stream of which it is a part.
- The **Distinguishing Bits** is a two bit field of which the first bit contains the value 1 and the second bit contains the value 0. It is used to distinguish EXI streams from text XML documents in the absence of an EXI cookie.
- **EXI Format Version** identifies the version of the EXI format being used. EXI format version numbers are integers.
- EXI Options provides a way to specify the options used to encode the body of the EXI stream (e.g. alignment, compression, schemaid, blcoksize, valueMaxLength)





- The rules for encoding a series of **Events** as an EXI stream are very simple and are driven by a declarative set of grammars that describes the structure of an EXI stream.
- Every event in the stream is encoded using the same set of encoding rules, which are summarized as follows:
 - 1. Get the next event data to be encoded
 - 2. If fidelity options indicate this event type is not processed, go to step 1
 - 3. Use the grammars to determine the *event code* of the event
 - 4. Encode the *event code* followed by the event content
 - 5. Evaluate the grammar production matched by the event
 - 6. Repeat until the *End Document* (ED) event is encoded



RDF Compression



Australian Government Department of Broadband, Communications and the Digital Economy

Australian Research Council

NICTA Funding and Supporting Members and Partners

Queensland Government





QUT





THE UNIVERSITY OF QUEENSLAND

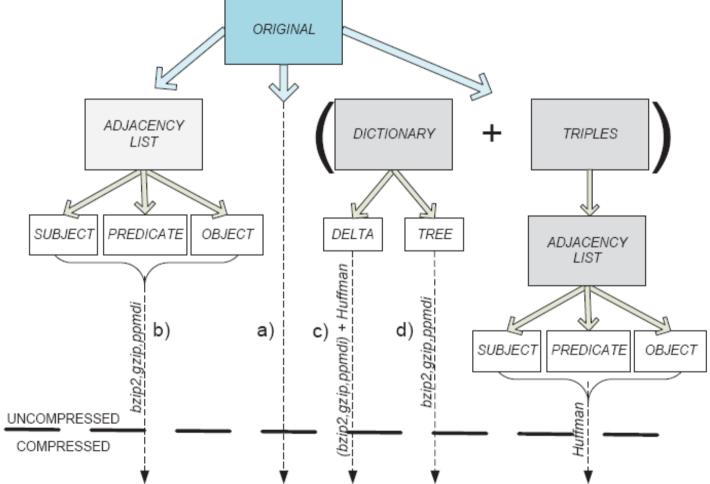
THE UNIVERSITY OF SYDNEY

Motivation of RDF Compression



- Large RDF datasets
- Verbosity and Redundancy.
- Examples
 - Billion Triple 2010 (~3200M triples, 318 gzipped chunks, ~27GB)
 - Uniprot (~845M, 12 gzipped chunks, ~23GB)





Javier D. Fernández, Claudio Gutierrez, Miguel A. Martinzez-Prieto. RDF compression: basic NICTA Copyright 2012 approaches. WWW, 2010



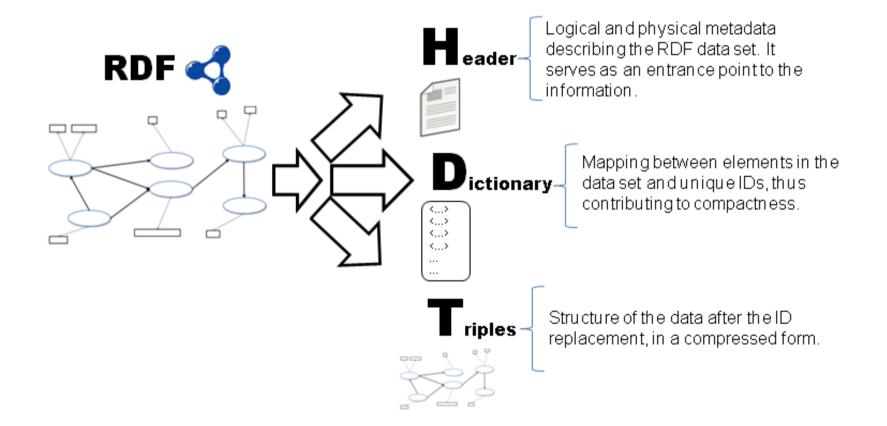
	B.Triples	Uniprot	US Census
Original Size	541.5	239.4	148.2
 a) Direct Compr. b) Adjacency List c) Delta+Triples d) Trie+Triples 	$\begin{array}{c} 37.9 \ (7.0\%) \ 35.5 \ (6.6\%) \ 37.9 \ (7.0\%) \ 39.4 \ (7.3\%) \end{array}$	$\begin{array}{c} 7.5 \ (3.1\%) \\ 5.3 \ (2.2\%) \\ 9.1 \ (3.8\%) \\ 9.7 \ (4.1\%) \end{array}$	$\begin{array}{c} 6.4 \ (4.3\%) \\ 4.8 \ (3.3\%) \\ 9.0 \ (6.1\%) \\ 9.0 \ (6.1\%) \end{array}$



- RDF data at big scale is highly compressible.
- Dedicated data structures, *e.g.* adjacency lists, code triples efficiently and facilitate compression (both string with ppmdi and integer with Huffman).
- RDF URIs are prone to efficient compression with standard techniques, but compression of literals deserve finer approaches.
- The structure of RDF graphs differs from XML or Web data, hence, classical approaches such as are not directly applicable.

HDT Overview



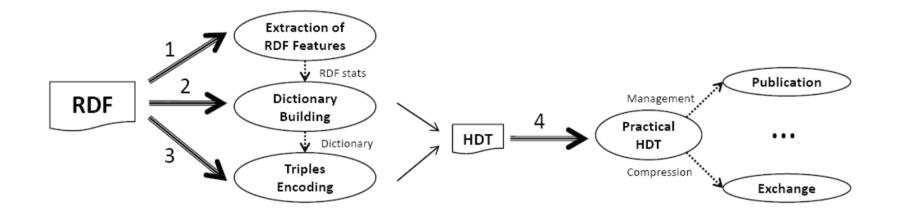


From imagination to impact

http://www.rdfhdt.org/

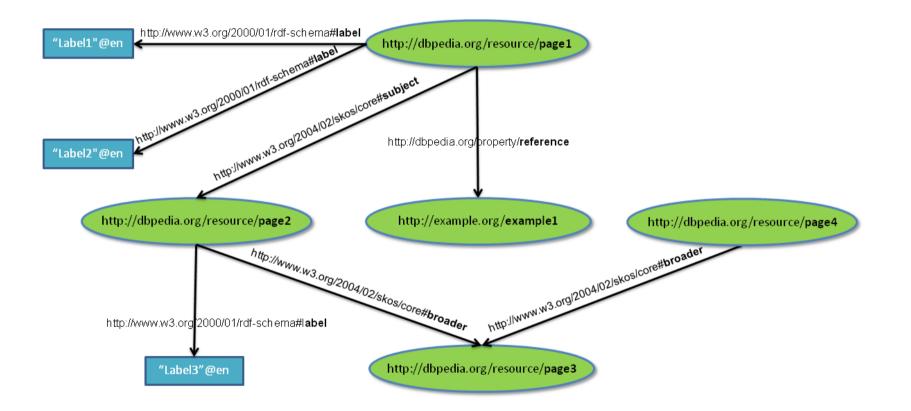
HDT Overview











DBPedia Header



<?xml version="1.0" encoding="UTF-8"?> <rdf:RDF xmlns:rdf="http://www.w3.org/1999/02/22-rdf-syntax-ns#" xmlns:hdt="http://purl.org/HDT/hdt#" xmlns:dc="http://purl.org/dc/terms/" xmlns:void="http://rdfs.org/ns/void#" xmlns:foaf="http://xmlns.com/foaf/0.1/" xmlns:scovo="http://purl.org/NET/scovo#" xmlns:rdfs="http://www.w3.org/2000/01/rdf-schema#" xmlns:swp="http://www.w3.org/2004/03/trix/swp-2/" xmlns:xsd="http://www.w3.org/2001/XMLSchema#"> <hdt:Dataset rdf:about="http://example.org/ex/DBpediaEN"> <hdt:publicationInformation rdf:parseType="Resource"> <void:dataDump rdf:resource="http://example.org/ex/DBpedia.rdf"/> <dc:title>DBpediaEN</dc:title> <dc:publisher> <foaf:Organization> <foaf:homepage rdf:resource="http://example.org/theCompany"/> </foaf:Organization> </dc:publisher> <dc:issued>2010-10-01</dc:issued> <dc:source rdf:resource="http://downloads.dbpedia.org/3.5.1/en/"/> <dc:license rdf:resource="http://www.gnu.org/copyleft/fdl.html"/> </hdt:publicationInformation> <hdt:statisticalInformation rdf:parseType="Resource"> <void:statItem rdf:parseType="Resource"> <scovo:dimension rdf:resource="void:numberOfTriples"/> <rdf:value>7</rdf:value> </void:statItem> <void:statItem rdf:parseType="Resource"> <scovo:dimension rdf:resource="hdt:subjectObjectRatio"/> <rdf:value>12.5</rdf:value> </void:statItem> </hdt:statisticalInformation>





```
<hdt:formatInformation rdf:parseType="Resource">

<!-- ... -->

</hdt:formatInformation>

<hdt:additionalInformation rdf:parseType="Resource">

<swp:signatureMethod rdf:resource="swp:JjcC14N-md5-xor-rsa"/>

<swp:signature rdf:datatype="xsd:base64Binary">AZ8QWE...</swp:signature>

</hdt:additionalInformation>

</hdt:additionalInformation>
```

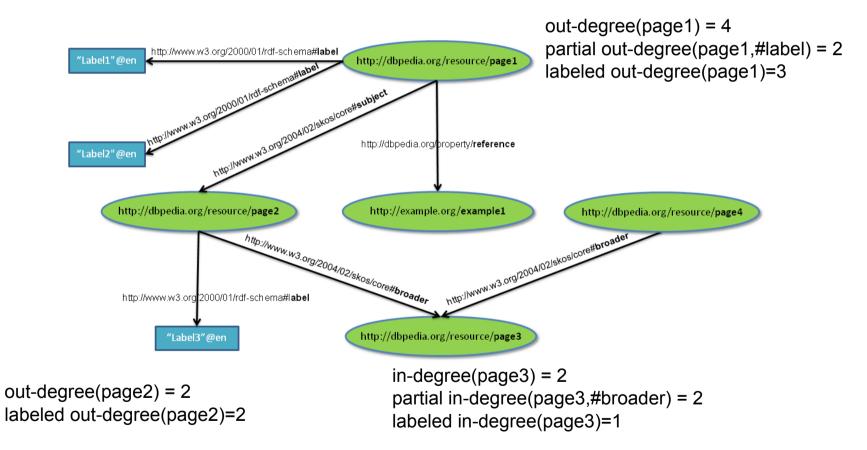
HDT Statistics



- out-degree, deg-(s)
 - the number of triples of G in which s occurs as subject
 - deg-(G), deg-(G)
- partial out-degree, deg- -(s, p)
 - the number of triples of G in which s occurs as subject and p as predicate
 - deg -(G), deg -(G)
- labeled out-degree, degL-(s)
 - the number of different predicates (labels) of G with which s is related as a subject in a triple of G
 - $degL^-(G), degL^-(G)$
- subject-object ratio, αs-o
 - the proportion of common subjects and objects in the graph G
 - $\alpha s o = |SG \cap OG| / |SG \cup OG|$
- Symmetrically, **in-degrees**: deg+(o), deg+(G), etc.







Dictionary in Practice

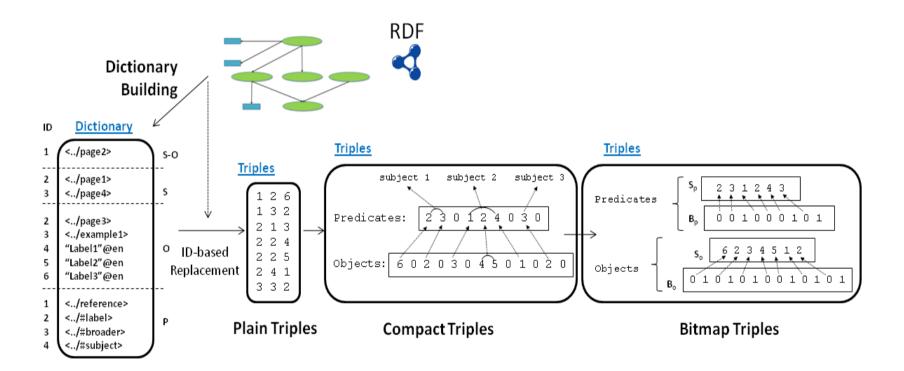


- Subset distinction:
 - (1) Common subject-objects
 - (2) The non common **subjects**
 - (3) The non common **objects**
 - (4) Predicates
- List of strings matching the mapping of the four subsets, in order from (1) to (4).
 - A reserved character is appended to the end of each string and each vocabulary to delimit their size.



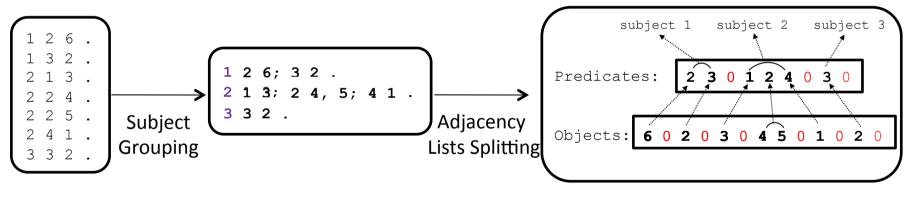


Contains the structure of the data after the ID replacement.



Compact Triples

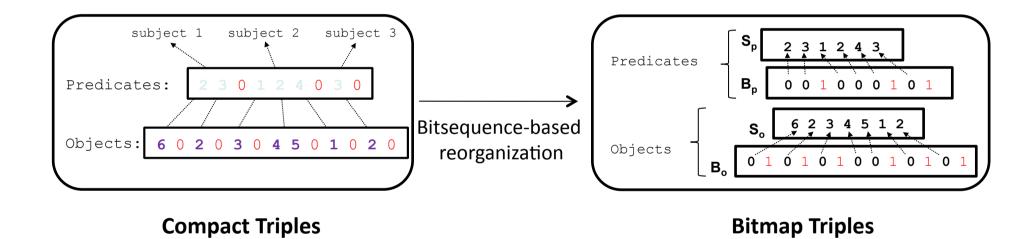




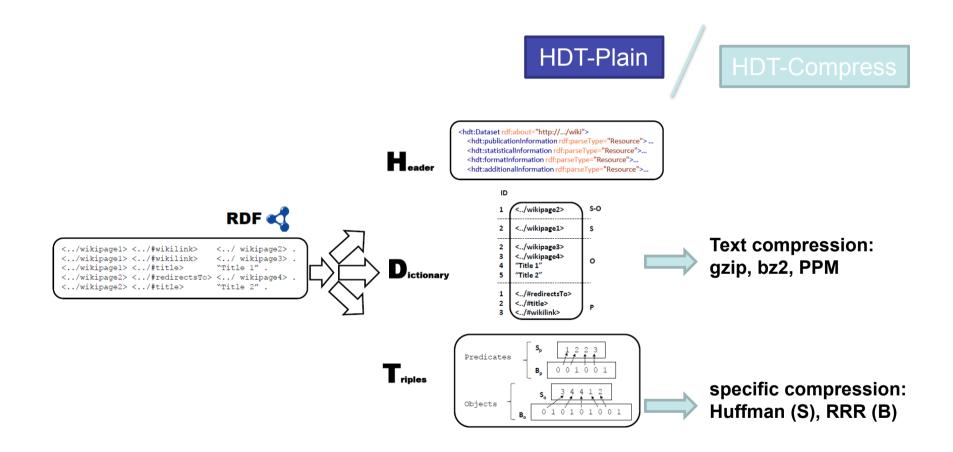
Compact Triples



Binary Triples



HDT Bitmap Triples Compression



Javier D. Fernández, Claudio Gutierrez, Miguel A. Martinzez-Prieto. : Compact Representation of Large RDF NICTA Copyright 2012 Data Sets for Publishing and Exchange. ISWC, 2010



RDF Compression



Australian Government Department of Broadband, Communications and the Digital Economy

Australian Research Council

NICTA Funding and Supporting Members and Partners

Queensland Government





QUT





THE UNIVERSITY OF QUEENSLAND

THE UNIVERSITY OF SYDNEY

Tutorial Overview

Session 1

XQuery Overview – Sherif SPARQL Overview – Axel XSPARQL: a combined language – Axel

Session 2 XQuery implementations – Sherif SPARQL implementations – Sherif XSPARQL implementations – Axel (optional) Compression formats for XML+RDF: EXI+HDT – Sherif

Q/A - Discussion

