







OLAP over Federated RDF Sources

DILSHOD IBRAGIMOV, KATJA HOSE, TORBEN BACH PEDERSEN, ESTEBAN ZIMÁNYI.

Outline

- Intro and Objectives
- Brief Intro to Technologies
- Our Approach and Progress
- Future Work

Business Intelligence and Semantic Web

 More and more data are published on the Web

 Business Intelligence tools need to analyze these data

• OnLine Analytical Processing (OLAP) style analysis of Linked Data may help in better decision making

• Expected challenges are data collection, integration, data aggregation...



Objectives

Design, develop, and evaluate an approach for performing OLAP over federated RDF sources

- Develop a framework for defining a multidimensional schema of a data cube expressed in RDF vocabularies in a global-as-view manner for further retrieval of data from different data sources
- Develop an approach for executing aggregate SPARQL queries over federated RDF sources
- Design an approach to support partial/result keeping materialization to store the results of previous requests and allow subsequent queries to execute faster
- Extend the query processing capabilities to include the inferred knowledge in materialized aggregate views of the linked data.

Language and Vocabularies

RDF (Resource Description Framework) is a standard model for data interchange on the Web

SPARQL is a query language for RDF. Queries defined in terms of graph patterns that are matched against the directed graph representing the RDF data

QB4OLAP is a special RDF vocabulary for OLAP Cubes on the Semantic Web.

VoID is an RDF Schema vocabulary for expressing metadata about RDF datasets

RDF

Statements about resources in the form of subject-predicate-object expressions

<P1> <is called> < Jimmy Wales >

TurtleExample : <P1> rdf:type foaf:Person .

<P1> foaf:name "Jimmy Wales" .

<P1> foaf:mbox <mailto:jwales@bomis.com> .

RDF extends the linking structure of the Web to use URIs to name the relationship between things

SPARQL Query Language

Developed by W3C Data Access Working Group

Queries defined in terms of graph patterns that are matched against the directed graph representing the RDF data

Ex: Show name and email of a person

PREFIX foaf: <http://xmlns.com/foaf/0.1/> SELECT ?name ?email WHERE {

?person a foaf:Person.?person foaf:name ?name.?person foaf:mbox ?email.

Name	Email
Jimmy Wales	mailto:jwales@bomis.com

SPARQL Query Language

4 types of queries to retrieve (read) data:

- SELECT
- CONSTRUCT
- ASK
- DESCRIBE

QB40LAP

QB4OLAP is a special RDF vocabulary for OLAP Cubes on the Semantic Web

-- Data structure definition and dimensions

exqb:NorthwindDW a qb:DataStructureDefinition ;

qb:component [qb:dimension exqb:Employee] ;

-- Definition of measures

qb:component [qb:measure exqb:Quantity].

-- Attributes

exqb4o:CompanyName a qb:AttributeProperty;

rdfs:comment "Company Name"@en .

-- Dimension

exqb4o:CustomerDim a rdf:Property, qb:DimensionProperty .

VoID

Vocabulary of Interlinked Datasets for expressing metadata about RDF datasets

- General metadata (following the Dublin Core model)
- Access metadata access to RDF data using various protocols
- Structural metadata structure and schema of datasets
- Description of links between datasets relation among multiple datasets

Towards Exploratory OLAP over Linked Open Data – A Case Study (Proposed System)



Source Discovery/Schema Builder is responsible for the discovery of data sources and construction of the Global Conceptual Schema

TOWARDS EXPLORATORY OLAP OVER LINKED OPEN DATA - A CASE STUDY (BIRTE 2014) 11

Proposed System



Global Conceptual Schema defines the high-level view of the system - expressed in QB4OLAP, VoID

TOWARDS EXPLORATORY OLAP OVER LINKED OPEN DATA - A CASE STUDY (BIRTE 2014) ¹²

Proposed System



Semantic Query Processor (using the Global Conceptual Schema) converts a user query to the new format and passes it to the Federated Query Processor

TOWARDS EXPLORATORY OLAP OVER LINKED OPEN DATA - A CASE STUDY (BIRTE 2014) ¹³

Proposed System



Federated Query Processor retrieves data from several federated data sources

TOWARDS EXPLORATORY OLAP OVER LINKED OPEN DATA - A CASE STUDY (BIRTE 2014) ¹⁴

Federated Query Processor – Motivating Example

- Hourly observation of radioactivity statistics at 47 prefectures
- Observations (March 16, 2011 March 15, 2012) converted to RDF data (places represented by URI from GeoNames)
- Interesting analyses:
 - AVG radioactivity separately for each prefecture in Japan
 - The MIN and MAX radioactivity for each prefecture (changes within one-year observations)

Motivating Example - Observation

```
#observation
<http://www.kanzaki.com/works/2011/
   stat/ra/20110414/p13/t08>
rdf:value "0.079"^^ms:microsv ;
ev:place <http://sws.geonames.org/
   1852083/> ;
ev:time <http://www.kanzaki.com/
   works/2011/stat/dim/d/
   20110414T08PT1H> ;
scv:dataset <http://www.kanzaki.com/
   works/2011/stat/ra/set/moe> .
#dimension - place
```

Motivating Example - Query

Ex: Show average radioactivity values for each prefecture

```
SELECT ?regName (AVG (?floatRV) AS ?average ) WHERE {
  ?s ev:place ?placeID ;
    ev:time ?time ;
    rdf:value ?radioValue .
  SERVICE <http://lod2.openlinksw.com/sparql> {
    ?placeID gn:parentFeature ?regionID . ?regionID gn:name ?regName .
  BIND (xsd:float (?radioValue ) as ?floatRV ).
GROUP BY ?regName
```

Motivating Example - Results

Virtuoso v07.10.3207, Sesame v2.7.11, and Jena Fuseki v1.0.0 (based on ARQ) timed out

- Network traffic analyzer showed that:
 - Virtuoso and Fuseki query GeoNames for each radioactivity observation (more than 400,000 requests)
 - Sesame is trying to download all triples that match the SERVICE query pattern (more than 7.8 million triples)

Basic Strategies - Mediator Join

- The mediator/federator receives the query from the ι
- The query optimizer sends separate queries to endpo and merges the results
- and merges the results
- Strong point parallelization
- Weak point expensive for large intermediate

```
results/datasets
```

```
SELECT ?placeID ?radioValue WHERE {
   ?s ev:place ?placeID; ev:time ?time.
   ?s rdf:value ?radioValue.
}
```



```
SELECT ?placeID ?regName WHERE {
    ?placeID gn:parentFeature ?regionID.
    ?regionID gn:name ?regName.
}
```

Basic Strategies - Semi-Join

• Main principle is to execute the subquery with the smallest result first and use the retrieved results as bindings for the join variable in other subqueries (SPARQL structure)

Efficient for highly selective subqueries (with FILTER statement)

SELECT ?regName (AVG (?radioValue) AS ?average) WHERE {

?s ev:place ?placeID . ?s ev:time ?time . ?s rdf:value ?radioValue .

SERVICE <http://lod2.openlinksw.com/sparql> {

?placeID gn: parentFeature ?regionID . ?regionID gn:name ?regName .

}

```
FILTER(?radioValue < 0.08).
```

} GROUP BY ?regName

Basic Strategies - Semi-Join (Cont)

SELECT ?placeID ?radioVal WHERE {

?s rdf:value ?radioVal ;

ev:place ?placeID; ev:time ?time.
FILTER (?radioValue < 0.08)</pre>

```
SELECT ?placeID ?regName
WHERE { ?placeID gn:parentFeature ?rgID.
    ?rgID gn:name ?regName.
    VALUES (?placeID) {
        <http://sws.geonames.org/1852083/>...}
```



Weak point - VALUES is not yet widely adopted in existing endpoints. SPARQL
 1.0 compliant alternatives of UNION (or FILTER) must often be used

Basic Strategies - Partial Aggregation

• If results are grouped by SERVICE query variables, further optimization is possible (motivating query example)

1) First group by the observation place (?placeID)

SELECT ?placeID (SUM (?floatRV) AS ?avgSUM) (COUNT (?floatRV) AS ?avgCNT) WHERE {

?s ev:place ?placeID . ?s ev:time ? time . ?s rdf:value ? radioValue .

BIND (xsd:float (?radioValue) as ?floatRV).

}

GROUP BY ?placeID

Basic Strategies - Partial Aggregation



• Final step – join the intermediate results and compute the final result (distributed/algebraic functions)

CODA – Cost-based Optimizer for Distributed Aggregate Queries

• **Decomposes** the original query into multiple subqueries (query Q_M and SERVICE queries $Q_{e1} \dots Q_{eN}$)

• Estimates query execution costs for different query execution plans

• Chooses the one with minimum costs

CODA - Costs

• Overall costs C_Q

$$C_Q = C_P + C_Q$$

• Communication costs C_C for subquery S_i :

 $C_C(S_i) = C_O + c_{S_i} * C_{map}$; C_O - communication establishing overhead, c_{S_i} - result size, and C_{map} - single result transfer cost

Processing costs

 $C_P = c_{agg_i} * C_{AGG}$; c_{agg_i} - number of aggregated observations, C_{AGG} - cost for processing a single observation

CODA - Estimating Constants

C_{map} - estimated using "SELECT * WHERE { ?s #p ?o . FILTER(?o = #o) } LIMIT #L"; different values for #L, #o and #p

 \circ C₀ - estimated with multiple "ASK {}" or "SELECT (1 AS ?v) {}"

C_{AGG} - estimated based on multiple "SELECT COUNT(?s) WHERE {?s ?p ?o }
 GROUP BY ?o"

Not perfectly accurate but the aim is to find out which execution plan is more efficient (not to predict the execution costs)

CODA - Result Size Estimation

Result size estimation - VoID statistics (dataset, property partition, class partition)

• c_t - total number of triples (void:triples), c_s - total number of distinct subjects (void:distinctSubjects), c_o - total number of distinct objects (void:distinctObjects)

• Single patterns - C_{res} for (?s ?p ?o) is given by c_t , (s ?p ?o) estimated as c_t/c_s , (?s ?p o) as c_t/c_o , and (s ?p o) as $c_t/(c_s * c_o)$; FILTER influence estimates

 Joins - estimates depend on shape (star vs path). Formulas taken from "Resource Planning for SPARQL Query Execution on Data Sharing Platforms".

type	pattern	cardinality card(join, partition)
subject – subject	?s predA ?o . ?s predB ?o2	$\frac{card(pat_1) \cdot card(pat_2)}{\max(cp_{predA,s}, cp_{predB,s})}$

CODA – Motivating Example

Decomposed into 3 queries

SELECT ?placeID (AVG(?floatRV) AS ?average) WHERE {

?s ev:place ?placeID .
?s rdf:value ?radioValue .
BIND(xsd:float(?radioValue) AS ?floatRV) .
?s ev:time ?time .

GROUP BY ?placeID

SELECT ?placeID ?floatRV WHERE

?s ev:place ?placeID .
?s rdf:value ?radioValue .
BIND(xsd:float(?radioValue) AS ?floatRV) .
?s ev:time ?time .

SELECT ?placeID ?regName WHERE

{

?placeID gn:parentFeature ?regionID .
?regionID gn:name ?regName .

CODA – Motivating Example

- Estimates for Radioact query:
 - number of aggregated triples: 405384
 - o estimated cost: 15
 - o number of returned triples: 405384
 - o estimated cost: 129
- Estimates for GeoNames query:
 - o number of returned triples: 7877627
 - estimated cost: 1956
- Selected plan Partial Aggregation

Test Case – SSB as RDF

• Star Schema Benchmark converted to RDF (strongly resembling SSB tabular structure)

• We generated data for different scale factors (1 to 5 - 6M to 30M observations, 110,5M to 547,5M triples)

- Different configurations
 - two endpoints (one endpoint containing main observation data and one SERVICE endpoint containing supporting data)
 - three endpoints (two SERVICE endpoints containing supporting data)
 - four endpoints (three SERVICE endpoints containing supporting data)
- All datasets and queries are available at http://extbi.cs.aau.dk/coda/

Aggregate Queries in Federations of Endpoints

- Efficiently processing aggregate queries in a federation of SPARQL endpoints
- Processing strategies (MedJoin, SemiJoin, PartialAgg)
- Cost-based Optimizer for Distributed Aggregate queries (CODA)
 - efficient and scalable
 - chooses the best query processing plan in different situations
 - significantly outperforms current state-of-the art triple stores

Improving Performance of Aggregate Queries

• With much data to process, analytical queries need special techniques to improve the performance of user queries

• One such technique is materializing the results of predefined queries and answering user queries based on these results – materialized views

Materializing RDF Views – Data Cube Example



Materializing RDF Views – Data Cube Example

• The total sales by month and customer state in 2010, for customers living in cities with over 1M inhabitants

SELECT ?c_state ?month (SUM(?total) AS ?sum_total) FROM <http://ex.com> WHERE {

?obs ex:OrderDate ?lo_orderdate ; ex:Customer ?customer ; ex:Revenue ?total .

?customer skos:broader ?c_city .

?c_city skos:broader ?c_state ; ex:population ?pop .

?lo_orderdate skos:broader ?month . ?month skos:broader ?year .

?year ex:value ?yearNum .

```
FILTER(?yearNum=2010 && ?pop > 1 000 000)
```

```
GROUP BY ?c_state ?month
```

Materializing RDF Views – Materializing RDF Data Cube

- Materializing all views in a data cube is not efficient
- Only several views with max benefit are chosen for materialization



Materializing RDF Views – Defining Views

• View query consists of 2 parts: *SELECT* query specifies the desired lattice node; *CONSTRUCT* query creates RDF triples from *SELECT* query results

CONSTRUCT {

?id ex:DateMonth ?vMonth ; ex:CustomerCity ?vCity ; ex:RevenueCount ?crev ; ex:RevenueSum ?srev .

WHERE {

SELECT ?id ?vCity ?vMonth (SUM(?rev) AS ?srev) (COUNT(?rev) AS ?crev)

WHERE {

?li ex:OrderDate ?odate ; ex:Customer ?cust ; ex:Revenue ?rev .

?cust skos:broader ?city .

?odate skos:broader ?vMonth .

BIND(IRI('http://ex.org/id#', CONCAT(?vCity, ?vMonth)) AS ?id).

GROUP BY ?id ?vState ?vMonth

Materializing RDF Views – Cost Model

• The cost of answering a query – number of triples contained in the materialized view used to answer the query

- Observation is described by its *n* dimensions and contains *m* measures.
- The total number of triples in a view (n + m) * N, where N is the number of observations

• In each step the algorithm selects a view with maximum benefit taking into account previously materialized views

Materializing RDF Views – Further Steps

• Answering Aggregate SPARQL Queries over Materialized Views with Inferred Knowledge

• Analyzing the Performance of Complex Aggregate SPARQL Queries with Intermediate Results Materialization

• Improving the Performance of OLAP Queries in a Federation of SPARQL Endpoints

Conclusion

• OnLine Analytical Processing (OLAP) style analysis of Linked Data may help in better decision making (e.g. analytics applications that integrate private data with web RDF datasets)

• The goal of the project is to improve the performance of analytical SPARQL queries over federated RDF sources