Ontology-Based Data Access with Ontop

Benjamin Cogrel benjamin.cogrel@unibz.it

KRDB Research Centre for Knowledge and Data Free University of Bozen-Bolzano, Italy



Freie Universität Bozen Libera Università di Bolzano Free University of Bozen-Bolzano

Optique

IT4BI, Blois, 22 April 2016

Ontology-Based Data Access (OBDA)

- SQL queries over tables can be hard to write manually
- 2 RDF and other Semantic Web standards
- Ontology-Based Data Access
- Optique platform
- Recent work
- 6 Conclusion



(1/40)

Outline

SQL queries over tables can be hard to write manually

- Toy example
- Industrial case: stratigraphic model design
- Semantic gap
- Solutions
- 2 RDF and other Semantic Web standards
- Ontology-Based Data Access
- Optique platform
- 6 Recent work
- Conclusion



Toy example: University Information System Relational source

uni1.student

<u>s_id</u>	first_name	last_name
1	Mary	Smith
2	John	Doe

uni1.academic

<u>a_id</u>	first_name	last_name	position
1	Anna	Chambers	1
2	Edward	May	9
3	Rachel	Ward	8

uni1.course

<u>c_id</u>	title
1234	Linear algebra

uni1.teaching

c_id	a₋id
1234	1
1234	2



(2/40)

Information need	SQL query
1. First and last names of the students	SELECT DISTINCT "first_name", "last_name" FROM "uni1"."student"
2. First and last names of the persons	SELECT DISTINCT "first_name", "last_name" FROM "uni1"."student" UNION SELECT DISTINCT "first_name", "last_name" FROM "uni1"."academic"
3. Course titles and teacher names	SELECT DISTINCT co."title", ac."last_name" FROM "uni1"."course" co, "uni1"."academic" ac, "uni1"."teaching" teach WHERE co."c_id" = teach."c_id" AND ac."a_id" = teach."a_id"
4. All the teachers	SELECT DISTINCT "a_id" FROM "uni1"."teaching" UNION SELECT DISTINCT "a_id" FROM "uni1"."academic" WHERE "position" BETWEEN 1 AND 8

Integration of a second source Fusion of two universities

uni2.person

<u>pid</u>	fname	Iname	status
1	Zak	Lane	8
2	Mattie	Moses	1
3	Céline	Mendez	2

uni2.course

<u>cid</u>	lecturer	lab_teacher	topic
1	1	3	Information
			security



Translation of information needs I

Information need	SQL query
1. First and last names of the students	SELECT DISTINCT "first_name", "last_name" FROM "uni1"."student" UNION SELECT DISTINCT "fname" AS "first_name",
2. First and last names of the persons	SELECT DISTINCT "first_name", "last_name" FROM "uni1"."student" UNION SELECT DISTINCT "first_name", "last_name" FROM "uni1"."academic" UNION SELECT DISTINCT "fname" AS "first_name",

Translation of information needs II

Information need	SQL query
3. Course titles and teacher names	SELECT DISTINCT co."title", ac."last_name" FROM "uni1"."course" co, "uni1"."academic" ac, "uni1"."teaching" teach WHERE co."c_id" = teach."c_id" AND ac."a_id" = teach."a_id" UNION SELECT DISTINCT co."topic" AS "title", pe."lname" AS "last_name" FROM "uni2"."person" pe, "uni2"."course" co WHERE pe."pid" = co."lecturer" OR pe."pid" = co."lab_teacher"

Translation of information needs III

Information need	SQL query
	SELECT DISTINCT 'uni1/' "a_id" AS "id" FROM "uni1"."teaching" UNION SELECT DISTINCT 'uni1/' "a_id" AS "id" FROM "uni1"."academic"
	WHERE "position" BETWEEN 1 AND 8 UNION
4. All the teachers	SELECT DISTINCT 'uni2/' "lecturer" AS "id" FROM "uni2"."course" UNION
	SELECT DISTINCT 'uni2/' "lab_teacher" AS "id" FROM "uni2"."course" UNION
	SELECT DISTINCT 'uni2/' "pid" AS "id" FROM "uni2"."person" WHERE "status" BETWEEN 6 AND 9

Industrial case: stratigraphic model design

Users: domain experts

- $\bullet \sim 900$ geologists et geophysicists
- Data collecting: 30-70% of their time



Sources

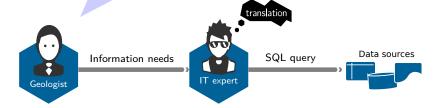
- ullet Exploitation and Production Data Store: ~ 1500 tables (100s GB)
- Norwegian Petroleum Directorate FactPages
- OpenWorks



(8/40)

Designing a new (ad-hoc) query

All norwegian wellbores of this type nearby this place having a permeability near this value. [...]
Attributes: completion date, depth, etc.



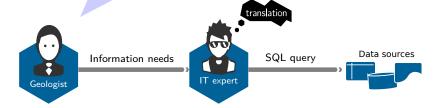
NB: Simplified information needs



(9/40)

Designing a new (ad-hoc) query

All norwegian wellbores of this type nearby this place having a permeability near this value. [...]
Attributes: completion date, depth, etc.



Takes 4 days in average (with EPDS only)

NB: Simplified information needs



(9/40)

Anonymized extract of a typical query

SELECT [...]

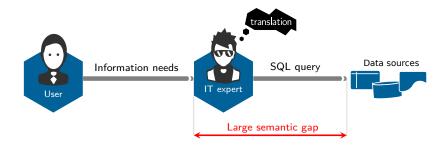
FROM db_name.table1 table1, db_name.table2 table2a, db name.table2 table2b. db_name.table3 table3a, db_name.table3 table3b, db name.table3 table3c. db name.table3 table3d. db_name.table4 table4a, db_name.table4 table4b, db name.table4 table4c. db name.table4 table4d. db_name.table4 table4e, db name.table4 table4f. db name.table5 table5a. db_name.table5 table5b, db name.table6 table6a. db name.table6 table6b. db name.table7 table7a. db_name.table7 table7b, db_name.table8 table8, db name.table9 table9. db name.table10 table10a. db_name.table10 table10b, db_name.table10 table10c, db_name.table11 table11, db_name.table12 table12, db name.table13 table13. db name.table14 table14. db_name.table15 table15, db name.table16 table16 WHERE [...]

table2a.attr1='keyword' AND table3a.attr2=table10c.attr1 AND table3a.attr6=table6a.attr3 AND table3a.attr9='keyword' AND table4a.attr10 IN ('keyword') AND table4a.attr1 IN ('keyword') AND table5a.kinds=table4a.attr13 AND table5b.kinds=table4c.attr74 AND table5b.name='keyword' AND (table6a.attr19=table10c.attr17 OR (table6a.attr2 IS NULL AND table10c.attr4 IS NULL)) AND table6a.attr14=table5b.attr14 AND table6a.attr2='kevword' AND (table6b.attr14=table10c.attr8 OR (table6b.attr4 IS NULL AND table10c.attr7 IS NULL)) AND table6b.attr19=table5a.attr55 AND table6b.attr2='keyword' AND table7a.attr19=table2b.attr19 AND table7a.attr17=table15.attr19 AND table4b.attr11='keyword' AND table8.attr19=table7a.attr80 AND table8.attr19=table13.attr20 AND table8.attr4='kevword' AND table9.attr10=table16.attr11 AND table3b.attr19=table10c.attr18 AND table3b.attr22=table12.attr63 AND table3b.attr66='keyword' AND table10a.attr54=table7a.attr8 AND table10a.attr70=table10c.attr10 AND table10a.attr16=table4d.attr11 AND table4c.attr99='keyword' AND table4c.attr1='keyword' AND

table11.attr10=table5a.attr10 AND table11.attr40='keyword' AND table11.attr50='kevword' AND table2b.attr1=table1.attr8 AND table2b.attr9 IN ('keyword') AND table2b.attr2 LIKE 'keyword'% AND table12.attr9 IN ('keyword') AND table7b.attr1=table2a.attr10 AND table3c.attr13=table10c.attr1 AND table3c.attr10=table6b.attr20 AND table3c.attr13='kevword' AND table10b.attr16=table10a.attr7 AND table10b.attr11=table7b.attr8 AND table10b.attr13=table4b.attr89 AND table13.attr1=table2b.attr10 AND table13.attr20=''keyword'' AND table13.attr15='keyword' AND table3d attr49mtable12 attr18 AND table3d.attr18=table10c.attr11 AND table3d.attr14='keyword' AND table4d.attr17 IN ('keyword') AND table4d.attr19 IN ('keyword') AND table16.attr28=table11.attr56 AND table16.attr16=table10b.attr78 AND table16.attr5=table14.attr56 AND table4e.attr34 IN ('keyword') AND table4e.attr48 IN ('keyword') AND table4f.attr89=table5b.attr7 AND table4f.attr45 IN ('keyword') AND table4f.attr1='keyword' AND table10c.attr2=table4e.attr19 AND (table10c.attr78=table12.attr56 OR (table10c.attr55 IS NULL AND table12.attr17 IS NULL))



Semantic gap



Querying over tables

Requires a lot of knowledge about:

- Magic numbers (e.g. $1 \rightarrow full\ professor$)
- Cardinalities and normal forms
- Spreading of closely-related information across many tables

Data integration

- Make things (much) worse!
- Variety: challenge #1 for most Big Data initiatives

unibz

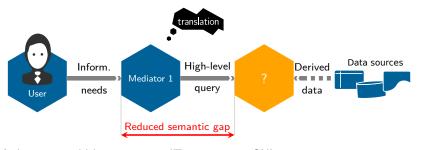
High-level translation

Main bottleneck: translation

- of the information needs
- ... into a formal query

Goal

Make such a translation easy (Ideally: IT expertise not required)



Mediator 1 could be a user, an IT expert or a GUI

General approach: two steps

- Translate the information needs into a high-level query
- Answer the high-level query automatically

Choice 1: How to derive data from the data sources

Extract Transform Load (ETL) process

E.g. relational data warehouse, triplestore



Virtual views

E.g. virtual databases (Teiid, Apache Drill, Exareme), OBDA (Ontop)



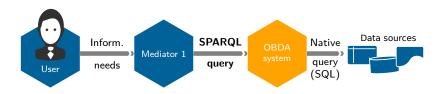


Choice 2: How to represent the derived data

New representation	Corresponding query language
Relational schema	SQL
JSON document	Mongo Aggregate, SQL (with e.g. Drill or Teiid)
XML document	XPath, XQuery, SQL (with e.g. Teiid)
RDF graph	SPARQL



Ontology-Based Data Access (OBDA)



Choice 1: How to derive data from the DBs

- Extract Transform Load (ETL) process
- Virtual views

Choice 2: How to represent the derived data

- New relational schema, JSON or XML documents
- Resource Description Framework (RDF)



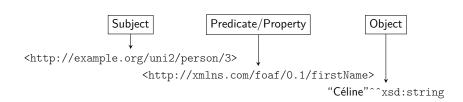
Outline

SQL queries over tables can be hard to write manually

- 2 RDF and other Semantic Web standards
 - RDF
 - SPARQL
 - Ontologies
 - Mappings
- Ontology-Based Data Access
- Optique platform
- 6 Recent work
- 6 Conclusion



Resource Description Framework (RDF)

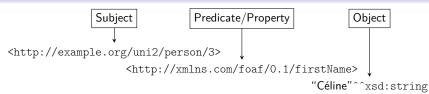




(16/40)

Semantic Web

Resource Description Framework (RDF) W3C standard

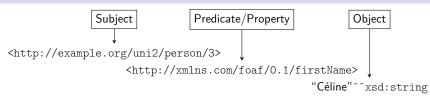


With the base IRI http://example.org/ and some prefixes:

<uni2/person/3> "Mendez"^^xsd:string foaf:lastName rdf:type <uni2/person/1> :AssociateProfessor :givesLecture <uni2/course/1> <uni2/person/1>



Resource Description Framework (RDF)



With the base IRI http://example.org/ and some prefixes:

Some characteristics

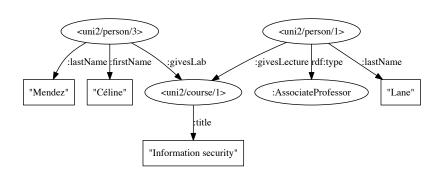
- Use global identifiers (IRI)
- Fixed arity (ternary)
- Self-descriptive
- Advanced: blank nodes

RDF graph

- Labelled directed graph
- Set of triples
- Trivial to merge
- Advanced: named graph

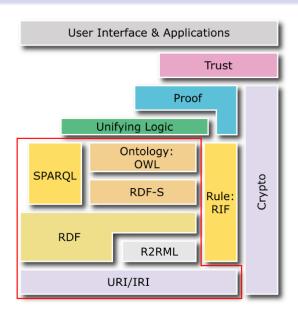


RDF graph





Semantic Web technologies Layer cake





(18/40)

SPARQL SPARQL Protocol and RDF Query Language

Title of courses taught by a professor and professor names

```
PREFIX : <http://example.org/voc#>
# Other prefixes omitted
SELECT ?title ?fName ?lName {
  ?teacher rdf:type :Professor .
  ?teacher :teaches ?course .
  ?teacher foaf:lastName ?lName .
  ?course :title ?title .
  OPTIONAL {
    ?teacher foaf:firstName ?fName .
```

Algebra

- Basic Graph Patterns
- OPTIONAL
- UNION
- GROUP BY
- MINUS
- FILTER NOT EXISTS



(19/40)

RDF Schema (RDFS)

Lightweight ontology

rdfs:subClassOf

rdfs:subPropertyOf

rdfs:domain

rdfs:range



Web Ontology Language (OWL)

Some constructs

owl:inverseOf

owl:disjointWith

owl:sameAs

Full OWL 2 is very expressive

- Many more constructs
- Computation costs become easily prohibitive

Profile OWL 2 QL Based on the Description Logic DL-Lite_R

Supported constructs

- Class and property hierarchies (rdfs:subClassOf and (rdfs:subPropertyOf)
- Property domain and range (rdfs:domain, rdfs:range)
- Inverse properties (owl:inverseOf)
- Class disjunction (owl:disjointWith)
- Mandatory participation (advanced)

Not supported

- Individual identities
 (owl:sameAs)
- Cardinality constraints (functional property, etc.)
- Many other constructs

Summary

- Lightweight ontologies
- A bit more than RDFS
- First-order rewritability (rewritable into a SQL query)



(22/40)

Mappings RDB-RDF

Ontop native format (similar to the R2RML standard)

Source (SQL)

```
SELECT s_id, firstName, lastName FROM uni1.student
```

Target (RDF, Turtle-like)

Result

- DBs unified into one RDF graph
- This graph can be queried with SPARQL



Mappings RDB-RDF Other mappings

Object property (:teaches)



Mappings RDB-RDF Other mappings

Object property (:teaches)

```
Target
        ex:uni1/academic/{a id} :teaches
(RDF)
                                    ex:uni1/course/{c id} .
        SELECT
Source
        FROM "uni1"."teaching"
```

Magic number

Target (RDF)	ex:uni1/academic/{a_id} a :FullProfessor .
Source	SELECT * FROM "uni1"."academic" WHERE "position" = 1



(24/40)

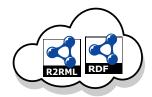
Outline

SQL queries over tables can be hard to write manually

- RDF and other Semantic Web standard
- 3 Ontology-Based Data Access
 - Querying the saturated RDF graph
 - Query reformulation
 - SQL query optimization
 - Ontop
- Optique platform
- Recent work
- 6 Conclusion



Querying the saturated RDF graph $_{\mbox{\scriptsize With SPARQL}}$

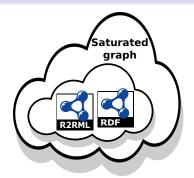




Querying the saturated RDF graph With SPARQL

Saturated RDF graph

- Saturation of the RDF graph derived from the mappings
- According to the ontology constraints
- Usually much bigger graph!

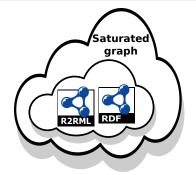




Querying the saturated RDF graph With SPARQL

Saturated RDF graph

- Saturation of the RDF graph derived from the mappings
- According to the ontology constraints
- Usually much bigger graph!



Materialized RDF graph

- ETL + saturation
- Maintenance
- + Expressive ontology profiles (like OWL 2 RL)
- + More SPARQL 1.1 constructs

Virtual RDF graph

- Query reformulation
- + No materialization
- Limited profiles like OWL 2 QL*



Query reformulation





(26/40)

L queries Semantic Web OBDA Optique platform Recent work Conclusion Reference:

Query reformulation



Role of the OWL 2 QL ontology

- Minor: SPARQL query rewriting (very specific cases)
- Main: mapping saturation (offline)



Query reformulation



Role of the OWL 2 QL ontology

- Minor: SPARQL query rewriting (very specific cases)
- Main: mapping saturation (offline)

Mapping saturation

- Query containment optimization
- Not only OWL 2 QL:
 - Horn fragment of OWL 2 [Botoeva et al., 2016c]
 - SWRL with linear recursion [Xiao et al., 2014]

L queries Semantic Web OBDA Optique platform Recent work Conclusion Reference

SQL query optimization

Objective : produce a SQL query. . .

- Similar to manually written ones
- Adapted to existing query planners



L queries Semantic Web OBDA Optique platform Recent work Conclusion Reference

SQL query optimization

Objective: produce a SQL query...

- Similar to manually written ones
- Adapted to existing query planners

Structural optimization

- From Join-of-unions to union-of-joins
- IRI decomposition to improve joining performance



QL queries Semantic Web OBDA Optique platform Recent work Conclusion Reference

SQL query optimization

Objective: produce a SQL query...

- Similar to manually written ones
- Adapted to existing query planners

Structural optimization

- From Join-of-unions to union-of-joins
- IRI decomposition to improve joining performance

Semantic optimization

- Redundant join elimination
- Redundant union elimination
- Using functional constraints



L queries Semantic Web OBDA Optique platform Recent work Conclusion References

SQL query optimization

Objective: produce a SQL query...

- Similar to manually written ones
- Adapted to existing query planners

Structural optimization

- From Join-of-unions to union-of-joins
- IRI decomposition to improve joining performance

Semantic optimization

- Redundant join elimination
- Redundant union elimination
- Using functional constraints

Functional constraints

- Primary and foreign keys, unique constraints
- Implicit in the business processes (Statoil)
- Vital for query reformulation!



(27/40)



Ontop framework

- Started in 2010
- Open-source (Apache 2)
- W3C standard compliant (SPARQL, OWL 2 QL, R2RML)
- Supports all major relational DBs (Oracle, DB2, Postgres, MySQL, etc.) and some virtual DBs (Teiid, Exareme)

Components

- Java APIs
- Protégé extension (GUI)
- Sesame endpoint

Integration

- Optique platform
- Stardog 4.0 (virtual graphs)

unidz

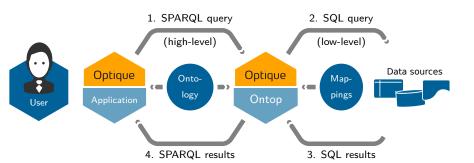
Outline

- SQL queries over tables can be hard to write manually
- 2 RDF and other Semantic Web standards
- Ontology-Based Data Access
- Optique platform
- 6 Recent work
- Conclusion



QL queries Semantic Web OBDA Optique platform Recent work Conclusion Reference:

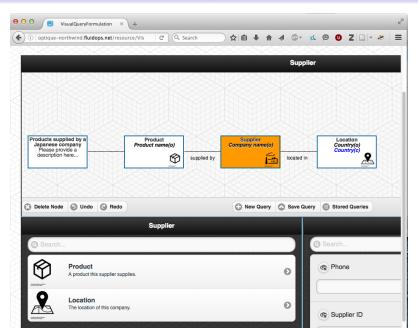
Optique platform





Visual query formulation (Optique VQS)

http://optique-northwind.fluidops.net demo/demo



Outline

- SQL queries over tables can be hard to write manually
- 2 RDF and other Semantic Web standards
- 3 Ontology-Based Data Access
- Optique platform
- Recent work
 - Cross-linked datasets
 - Beyond OWL 2 QL
 - MongoDB support
- 6 Conclusion



Cross-linked datasets [Calvanese et al., 2015]

Linking tables

- Different identifiers used across datasets
- Tables keeping track of the equivalence

 D_1 and D_2

id1	id2
a1	b2
a2	b1

 D_2 and D_3

id2	id3
b1	c4
b2	сЗ

 D_1 and D_3

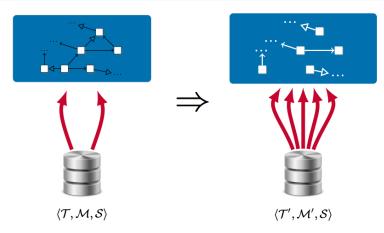
id1	id3
a3	c5

Support for linking tables

- SPARQL query rewriting
- owl:sameAs properties specified in the mappings
- Pruning based on incompatible URI templates

Beyond OWL 2 QL (I)

Framework for rewriting and approximation of OBDA specifications [Botoeva et al., 2016c]



Rewriting The new specification is equivalent to the original one w.r.t. query answering (query-inseparable).

Approximation The new specification is a sound approximation of the original one w.r.t. query answering.



Beyond OWL 2 QL (II)

```
 \begin{array}{|c|c|c|}\hline \mathcal{T} = \{ & A \sqcap B \sqsubseteq C \\ \mathcal{M} = \{ & \mathsf{SQL}_A(x) \leadsto A(x), \\ & \mathsf{SQL}_B(x) \leadsto B(x) \\ \end{array} \} \Rightarrow \begin{array}{|c|c|c|c|}\hline \mathcal{H}' = \{ & \mathsf{SQL}_A(x) \leadsto A(x), \\ & \mathsf{SQL}_B(x) \leadsto B(x), \\ & \mathsf{SQL}_A(x) \land \mathsf{SQL}_B(x) \\ \hline \end{array}
```

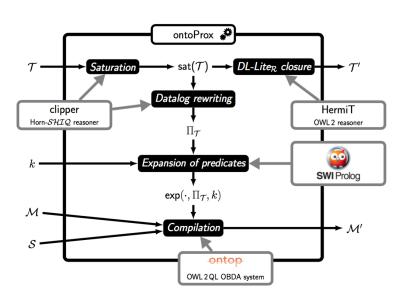
$$\begin{split} \mathcal{T}' &= \{ \ \} \\ \mathcal{M}' &= \{ \ \mathsf{SQL}_A(x) \leadsto A(x), \\ \mathsf{SQL}_R(x,y) \leadsto R(x,y), \\ \mathsf{SQL}_R(x,y) \land \mathsf{SQL}_A(y) \leadsto A(x) \\ \mathsf{SQL}_R(x,y) \land \mathsf{SQL}_R(y,z) \land \mathsf{SQL}_A(z) \leadsto A(x) \\ \mathsf{SQL}_R(x,y) \land \mathsf{SQL}_R(y,z) \land \mathsf{SQL}_R(z,w) \land \mathsf{SQL}_A(w) \leadsto A(x) \ \} \end{split}$$



s Semantic Web OBDA Optique platform Recent work Conclusion References

Beyond OWL 2 QL (III)

New tool: ontoprox





Mongo DB

A popular document database

```
JSON document described an awarded scientist
```

Persons who received two awards in the same year

Semantic Web OBDA Optique platform Recent work Conclusion Reference

MongoDB support

[Botoeva et al., 2016a] [Botoeva et al., 2016b]

MongoDB

- Document database
- JSON-like documents
- Does not respect first normal form (arrays)

Mongo Aggregation Framework

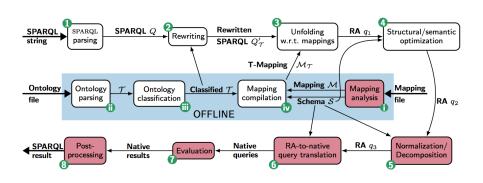
- Query language
- Use absolute paths
- At least as expressive as relational algebra (MUPGL fragment)

Integration in the OBDA setting

- JSON-RDF mapping language
- (First normal form) relational views over MongoDB
- Translation from relational algebra

eries Semantic Web OBDA Optique platform Recent work Conclusion References

Evolution of the Ontop architecture For supporting non-relational databases



In red: components that are DB-specific.



Outline

- SQL queries over tables can be hard to write manually
- 2 RDF and other Semantic Web standards
- Ontology-Based Data Access
- Optique platform
- Recent work
- 6 Conclusion



(38/40)

QL queries Semantic Web OBDA Optique platform Recent work Conclusion Reference:

Conclusion

Main message: we need high-level access to data

- SQL queries over tables can be difficult to write manually (low-level)
- OBDA is a powerful solution for high-level data access
- Ontop is an open-source OBDA framework

Work in progress

- SPARQL aggregation
- MongoDB
- Better SPARQL OPTIONAL
- SPARQL MINUS

Links

- Github : ontop/ontop
- ontop4obda@googlegroups.com
- Twitter : @ontop4obda
- http://ontop.inf.unibz.it
- http://optique-project.eu



queries Semantic Web OBDA Optique platform Recent work Conclusion Reference

Ontop team

- Diego Calvanese
- Guohui Xiao
- Elena Botoeva
- Roman Kontchakov (Birbeck, London)
- Sarah Komla-Ebri
- Elem Güzel Kalayci
- Ugur Dönmez
- Davide Lanti
- Dag Hovland (Oslo)
- Mariano Rodriguez-Muro (now in IBM Research, NY)
- Martin Rezk (now in Rakuten, Tokyo)
- Me



L queries Semantic Web OBDA Optique platform Recent work Conclusion References

Introductory resources

Journal paper [Calvanese et al., 2016]

Ontop: Answering SPARQL Queries over Relational Databases.

Diego Calvanese, Benjamin Cogrel, Sarah Komla-Ebri, Roman Kontchakov,
Davide Lanti, Martin Rezk, Mariano Rodriguez-Muro, and Guohui Xiao.

Semantic Web Journal. 2016 http://www.semantic-web-journal.net/content/ontop-answering-sparql-queries-over-relational-databases-1

Tutorial

https://github.com/ontop/ontop-examples/tree/master/swj-2015

University example

https://github.com/ontop/ontop-examples/tree/master/university

EPNET SPARQL endpoint

http://136.243.8.213/epnet-pleiades-edh/

References I

[Botoeva et al., 2016a] Elena Botoeva, Diego Calvanese, Benjamin Cogrel, Martin Rezk, and Guohui Xiao.

A formal presentation of MongoDB (Extended version).

CoRR Technical Report abs/1603.09291, arXiv.org e-Print archive, 2016.

Available at http://arxiv.org/abs/1603.09291.

[Botoeva et al., 2016b] Elena Botoeva, Diego Calvanese, Benjamin Cogrel, Martin Rezk, and Guohui Xiao.

OBDA beyond relational DBs: A study for MongoDB.

In International workshop on Description Logic, 2016.

[Botoeva *et al.*, 2016c] Elena Botoeva, Diego Calvanese, Valerio Santarelli, Domenico F. Savo, Alessandro Solimando, and Guohui Xiao.

Beyond OWL 2 QL in OBDA: Rewritings and approximations.

In Proc. of the 30th AAAI Conf. on Artificial Intelligence (AAAI), 2016.

[Calvanese et al., 2015] Diego Calvanese, Martin Giese, Dag Hovland, and Martin Rezk.

Ontology-based integration of cross-linked datasets.

volume 9366 of LNCS, pages 199-216. Springer, 2015.



References II

[Calvanese et al., 2016] Diego Calvanese, Benjamin Cogrel, Sarah Komla-Ebri, Roman Kontchakov, Davide Lanti, Martin Rezk, Mariano Rodriguez-Muro, and Guohui Xiao.

Ontop: Answering SPARQL queries over relational databases.

Semantic Web J., 2016.

DOI: 10.3233/SW-160217.

[Xiao et al., 2014] Guohui Xiao, Martin Rezk, Mariano Rodriguez-Muro, and Diego Calvanese.

Rules and ontology based data access.

volume 8741 of *LNCS*, pages 157-172. Springer, 2014.

