

Semantic Technologies for Data Access and Integration

Part 3: Query Processing and Optimization

Diego Calvanese, Guohui Xiao

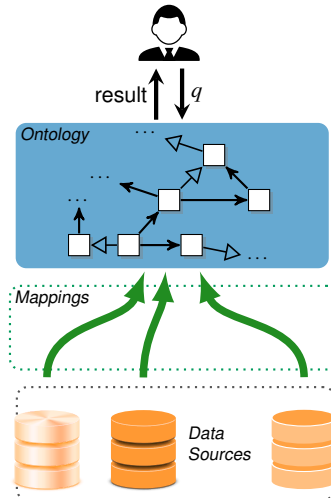
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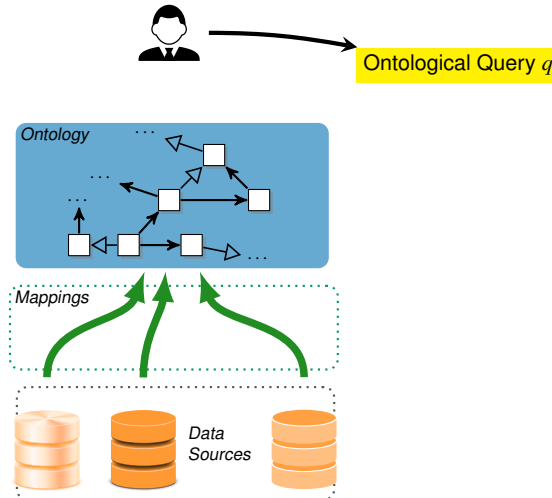
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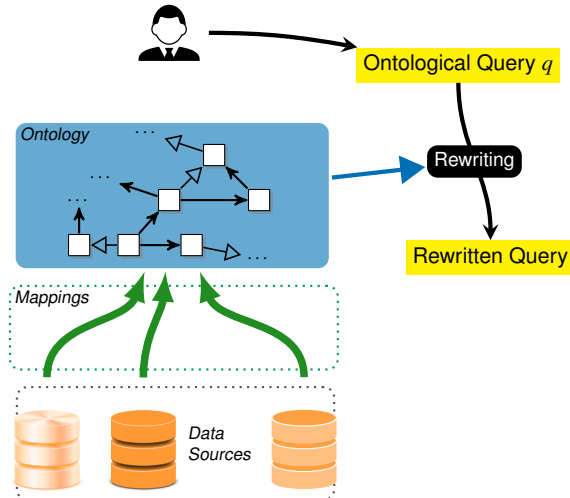
Query answering by query rewriting



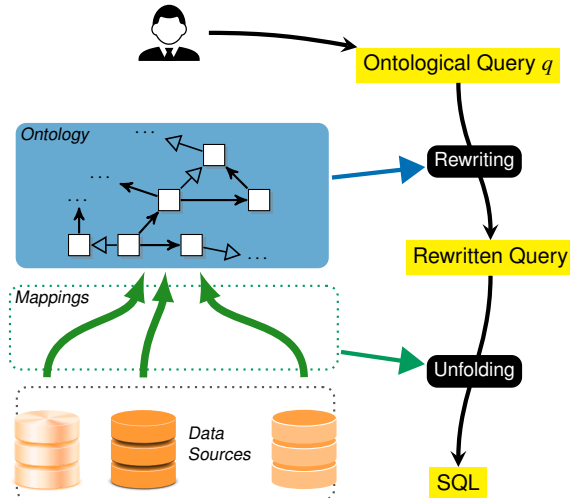
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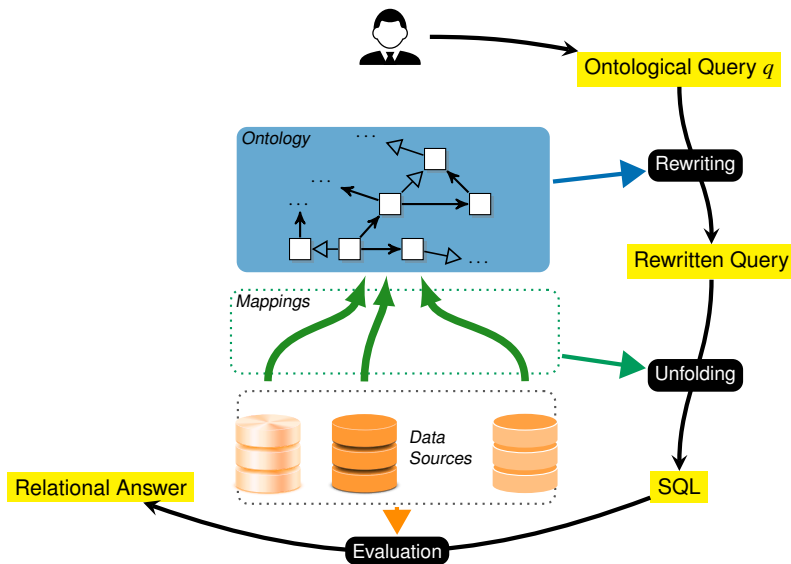
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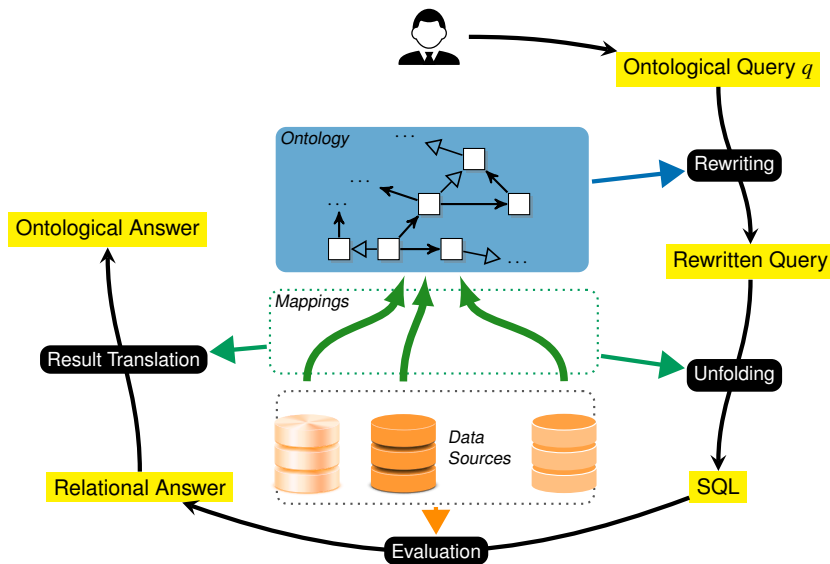
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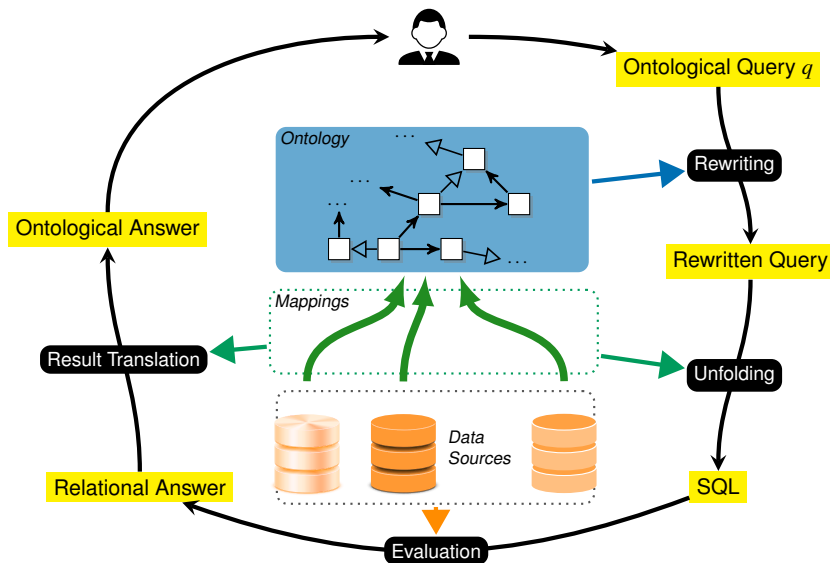
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Query answering by query rewriting



Outline

- 1 Query rewriting wrt an OWL 2 QL ontology
- 2 Saturation and optimization of the mapping
- 3 Query reformulation and optimization

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Query answering via query reformulation

To compute the certain answers to a SPARQL query q over an OBDA instance $O = \langle \mathcal{P}, \mathcal{D} \rangle$, with $\mathcal{P} = \langle \mathcal{T}, \mathcal{S}, \mathcal{M} \rangle$:

- 1 Compute the perfect rewriting of q w.r.t. \mathcal{T} .
- 2 Unfold the perfect rewriting wrt the mapping \mathcal{M} .
- 3 Optimize the unfolded query, using database constraints.
- 4 Evaluate the resulting SQL query over \mathcal{D} .

Steps 1 – 3 are collectively called **query reformulation**.

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The rewriting Step 1 deals with the objects that are existentially implied by the axioms of the ontology.

Example of existential reasoning

Suppose that every graduate student is supervised by some professor, i.e.

GraduateStudent $\sqsubseteq \exists \text{isSupervisedBy}.\text{Professor}$

and john is a graduate student: *GraduateStudent(john)*.

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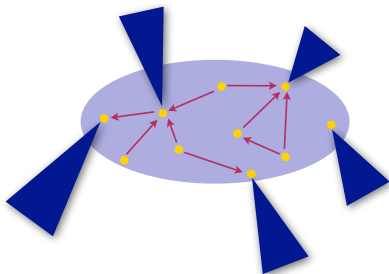
$q(x) \leftarrow isSupervisedBy(x, y), Professor(y)$

The answer should be **john**, even though we don't know who is John's supervisor (under existential reasoning).

Existential reasoning and query rewriting

Canonical model

Every consistent *DL-Lite* KB $\mathcal{K} = (\mathcal{T}, \mathcal{A})$ has a **canonical model** $\mathcal{I}_{\mathcal{K}}$, which **gives the right answers to all CQs**, i.e., $\text{cert}(q, \mathcal{K}) = \text{ans}(q, \mathcal{I}_{\mathcal{K}})$

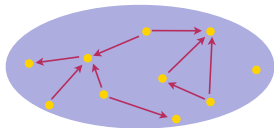


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Core
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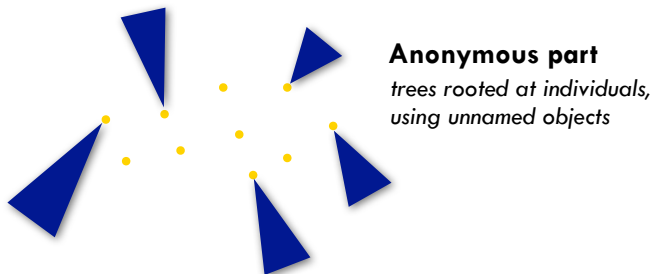


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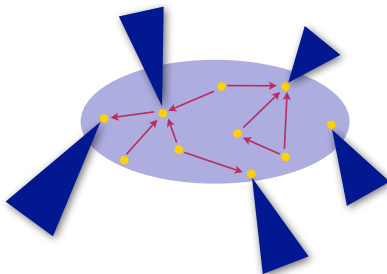


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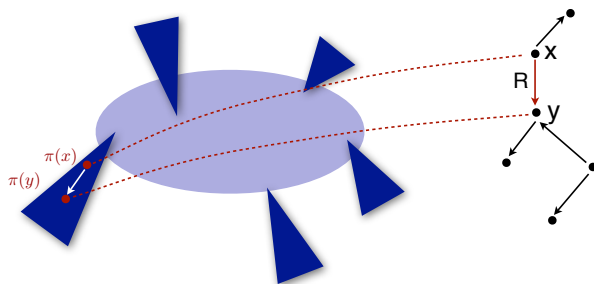


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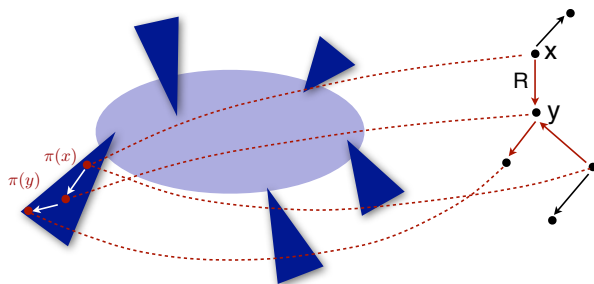


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Example of existential reasoning (continued)

Using the (tree witness) rewriting algorithm, the query

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is rewritten to a union of two conjunctive queries (or a SPARQL union query):

$$\begin{aligned} q(x) &\leftarrow \textit{isSupervisedBy}(x, y), \textit{Professor}(y) \\ q(x) &\leftarrow \textit{GraduateStudent}(x) \end{aligned}$$

Therefore, over the Abox *GraduateStudent(john)*, the rewritten query returns *john* as an answer.

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Note: In *Ontop*, if one wants to answer queries by performing existential reasoning, the tree-witness rewriting algorithm needs to be switched on explicitly.

The *PerfectRef* algorithm for query rewriting

To illustrate Step ① of the query reformulation algorithm, we briefly describe *PerfectRef*, a simple query rewriting algorithm that requires to iterate over:

- rewriting steps that involve TBox inclusion assertions, and
- unification of query atoms.

The perfect rewriting of q is still a SPARQL query involving UNION.

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- rewriting steps that involve TBox inclusion assertions, and
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The perfect rewriting of q is still a SPARQL query involving UNION.

Note: disjointness assertions play a role in ontology satisfiability, but can be ignored during query rewriting (i.e., we have **separability**).

Query rewriting step: Basic idea

Intuition: an **inclusion assertion** corresponds to a **logic programming rule**.

Example

The inclusion assertion $FullProf \sqsubseteq Prof$
corresponds to the logic programming rule $Prof(z) \leftarrow FullProf(z).$

Query rewriting step: Basic idea

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Basic rewriting step:

When an atom in the query unifies with the **head** of the rule, generate a new query by substituting the atom with the **body** of the rule.

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Consider the query $q(x) \leftarrow Prof(x)$.

By applying the inclusion assertion to the atom $Prof(x)$, we generate:

$q(x) \leftarrow FullProf(x)$.

This query is added to the input query, and contributes to the perfect rewriting.

Query rewriting (cont'd)

Example

Consider the query $q(x) \leftarrow \text{teaches}(x, y), \text{Course}(y)$

and the inclusion assertion $\exists \text{teaches}^- \sqsubseteq \text{Course}$

as a logic programming rule: $\text{Course}(z_2) \leftarrow \text{teaches}(z_1, z_2).$

The inclusion applies to $\text{Course}(y)$, and we add to the rewriting the query

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as a logic programming rule: $\text{teaches}(z, f(z)) \leftarrow \text{FullProf}(z).$

The inclusion applies to $\text{teaches}(x, y)$, and we add to the rewriting the query

$$q(x) \leftarrow \text{FullProf}(x).$$

Query rewriting – Constants

Example

Conversely, for the query $q(x) \leftarrow teaches(x, databases)$

and the same inclusion assertion as before

$$FullProf \sqsubseteq \exists teaches$$

as a logic programming rule:

$$teaches(z, f(z)) \leftarrow FullProf(z)$$

$teaches(x, databases)$ does not unify with $teaches(z, f(z))$, since the **skolem term** $f(z)$ in the head of the rule **does not unify** with the constant $databases$.

Remember: We adopt the **unique name assumption**.

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Example

The same holds for the following query, where y is **distinguished**, since unifying $f(z)$ with y would correspond to returning a skolem term as answer to the query:

$$q(x, y) \leftarrow teaches(x, y).$$

Query rewriting – Join variables

An analogous behavior to the one with constants and with distinguished variables holds when the atom contains **join variables** that would have to be unified with skolem terms.

Example

Consider the query $q(x) \leftarrow \text{teaches}(x, y), \text{Course}(y)$

and the inclusion assertion $\text{FullProf} \sqsubseteq \exists \text{teaches}$

as a logic programming rule: $\text{teaches}(z, f(z)) \leftarrow \text{FullProf}(z)$.

The inclusion assertion above does **not** apply to the atom $\text{teaches}(x, y)$.

Query rewriting – Reduce step

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Example

However, we can transform the above query by **unifying** the atoms $\text{teaches}(x, y)$ and $\text{teaches}(z, y)$. This rewriting step is called **reduce**, and produces the query

$$q(x) \leftarrow \text{teaches}(x, y).$$

Now, we can apply the inclusion above, and add to the rewriting the query

$$q(x) \leftarrow \text{FullProf}(x).$$

Query rewriting – Summary

To **compute the perfect rewriting** of a query q , start from q , iteratively get a CQ q' to be processed, and do one of the following:

- **Apply** to some atom of q' an **inclusion assertion** in \mathcal{T} as follows:

| | | | |
|---------------------------------------|----------------------------|--------------------|----------------------------|
| $A_1 \sqsubseteq A_2$ | $\dots, A_2(x), \dots$ | \rightsquigarrow | $\dots, A_1(x), \dots$ |
| $\exists P \sqsubseteq A$ | $\dots, A(x), \dots$ | \rightsquigarrow | $\dots, P(x, _), \dots$ |
| $\exists P^- \sqsubseteq A$ | $\dots, A(x), \dots$ | \rightsquigarrow | $\dots, P(_, x), \dots$ |
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| $P_1 \sqsubseteq P_2$ | $\dots, P_2(x, y), \dots$ | \rightsquigarrow | $\dots, P_1(x, y), \dots$ |
| $P_1 \sqsubseteq P_2^-$ | $\dots, P_2(x, y), \dots$ | \rightsquigarrow | $\dots, P_1(y, x), \dots$ |

('_' denotes a variable that appears only once)

- Choose two atoms of q' that unify, and **apply the unifier** to q' .

Each time, the result of the above step is added to the queries to be processed.

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The UCQ resulting from this process is the **perfect rewriting** $r_{q, \mathcal{T}}$.

Query rewriting algorithm

Algorithm *PerfectRef*(Q, \mathcal{T}_P)

Input: union of conjunctive queries Q , set \mathcal{T}_P of *DL-Lite* inclusion assertions

Output: union of conjunctive queries PR

$PR := Q$;

repeat

$PR' := PR$;

for each $q \in PR'$ **do**

for each g in q **do**

for each inclusion assertion I in \mathcal{T}_P **do**

if I is applicable to g **then** $PR := PR \cup \{ \text{ApplyPI}(q, g, I) \}$;

for each g_1, g_2 in q **do**

if g_1 and g_2 unify **then** $PR := PR \cup \{ \tau(\text{Reduce}(q, g_1, g_2)) \}$;

until $PR' = PR$;

return PR

Observations:

- Termination follows from having only finitely many different rewritings.
- Disjointness assertions and functionalities do not play any role in the rewriting of the query.

Query answering in *DL-Lite* – Example

TBox:

$FullProf \sqsubseteq Prof$

$Prof \sqsubseteq \exists teaches$

$\exists teaches^- \sqsubseteq Course$

Corresponding rules:

$Prof(x) \leftarrow FullProf(x)$

$\exists y(teaches(x, y)) \leftarrow Prof(x)$

$Course(x) \leftarrow teaches(y, x)$

Query: $q(x) \leftarrow teaches(x, y), Course(y)$

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$q(x) \leftarrow teaches(x, -)$

$q(x) \leftarrow Prof(x)$

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TBox:

$FullProf \sqsubseteq Prof$

$Prof \sqsubseteq \exists teaches$

$\exists teaches^- \sqsubseteq Course$

Corresponding rules:

$Prof(x) \leftarrow FullProf(x)$

$\exists y(teaches(x, y)) \leftarrow Prof(x)$

$Course(x) \leftarrow teaches(y, x)$

Query: $q(x) \leftarrow teaches(x, y), Course(y)$

Perfect rewriting: $q(x) \leftarrow teaches(x, y), Course(y)$

$q(x) \leftarrow teaches(x, y), teaches(-, y)$

$q(x) \leftarrow teaches(x, -)$

$q(x) \leftarrow Prof(x)$

$q(x) \leftarrow FullProf(x)$

Query answering in *DL-Lite* – Example

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$q(x) \leftarrow teaches(x, .)$

$q(x) \leftarrow Prof(x)$

$q(x) \leftarrow FullProf(x)$

ABox: $teaches(jim, databases) \quad FullProf(jim)$
 $teaches(julia, security) \quad FullProf(nicole)$

Evaluating the perfect rewriting over the ABox (seen as a DB) produces as answer
{jim, julia, nicole}.

Query answering in *DL-Lite* – An interesting example

TBox: $Person \sqsubseteq \exists hasFather$
 $\exists hasFather^- \sqsubseteq Person$

ABox: $Person(john)$

Query: $q(x) \leftarrow Person(x), hasFather(x, y_1), hasFather(y_1, y_2), hasFather(y_2, y_3)$

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\Downarrow **Apply** $Person \sqsubseteq \exists hasFather$ to the atom $hasFather(y_2, -)$

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Complexity of query answering in *DL-Lite*

Query answering for UCQs / SPARQL queries is:

- Efficiently tractable in the size of the **TBox**, i.e., **P**TIME.
- Very efficiently tractable in the size of the **ABox**, i.e., **AC**⁰.
- Exponential in the size of the **query**, more precisely **NP-complete**.

In **theory this is not bad**, since this is precisely the complexity of evaluating CQs in plain relational DBs.

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Can we go beyond *DL-Lite*?

Essentially no! By adding essentially any additional DL constructor we lose first-order rewritability and hence these nice computational properties.

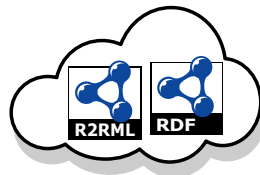
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Querying the OBDA system

OBDA system $\mathcal{K} = \langle \mathcal{T}, \mathcal{M}, \mathcal{D} \rangle$

- $DL\text{-}Lite_{\mathcal{R}}$ TBox \mathcal{T}
- RDF graph \mathcal{G} obtained from the mapping \mathcal{M} and the data sources \mathcal{D}
- \mathcal{G} can be viewed as the ABox



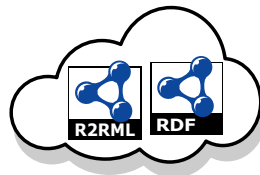
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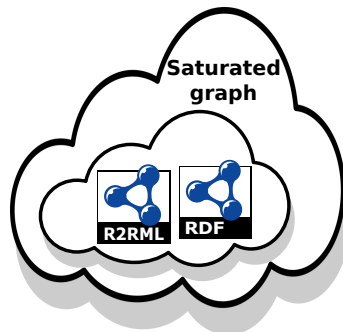
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Saturated RDF graph \mathcal{G}_{sat}

- Saturation of \mathcal{G} w.r.t. \mathcal{T}
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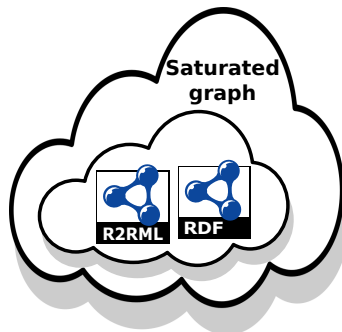
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Query answering

- SPARQL query q over \mathcal{K}
- If there is no existential restriction $B \sqsubseteq \exists R.C$ in \mathcal{T} , q can be directly evaluated over \mathcal{G}_{sat}



Saturated RDF graph \mathcal{G}_{sat}

- Saturation of \mathcal{G} w.r.t. \mathcal{T}
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How to handle the RDF graph \mathcal{G}_{sat} in practice?

By materializing it

- Materialization of \mathcal{G} (ETL)
+ saturation
- Large volume
- Maintenance
- Typical profile: OWL 2 RL

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By keeping it virtual

- Query rewriting
- + No materialization required
- Saturated mapping \mathcal{M}_{sat}
- Typical profile: OWL 2 QL

H-complete ABox

[Rodriguez-Muro, Kontchakov, and Zakharyashev 2013; Kontchakov and Zakharyashev 2014]

ABox saturation

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- Let \mathcal{K} be a $DL\text{-}Lite_{\mathcal{R}}$ knowledge base, and let \mathcal{K}_{sat} be the result of saturating \mathcal{K} . Then, for every ABox assertion α , we have:

$$\mathcal{K} \models \alpha \quad \text{iff} \quad \alpha \in \mathcal{K}_{\text{sat}}$$

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Saturated mapping \mathcal{M}_{sat} (also called *T-mapping*)

- Composition of the mapping \mathcal{M} and the $DL\text{-}Lite_{\mathcal{R}}$ TBox \mathcal{T} .
- \mathcal{M}_{sat} applied to \mathcal{D} produces \mathcal{G}_{sat} (H-complete ABox).

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- \mathcal{M}_{sat} applied to \mathcal{D} produces \mathcal{G}_{sat} (H-complete ABox).
- Does not depend of the SPARQL query q (can be pre-computed).
- Can be optimized (exploiting query containment).

TBox, user-defined mapping assertions, and foreign key

 $Student \sqsubseteq Person$ $Student(iri1(scode)) \Leftarrow student(scode, fn, ln) \quad (1)$ $PostDoc \sqsubseteq Person$ $PostDoc(iri2(icode)) \Leftarrow academic(icode, fn, ln, pos), pos = 9 \quad (2)$ $AssociateProfessor \sqsubseteq Person$ $AssociateProfessor(iri2(icode)) \Leftarrow academic(icode, fn, ln, pos), pos = 2 \quad (3)$ $\exists teaches \sqsubseteq Person$ $FacultyMember(iri2(icode)) \Leftarrow academic(icode, fn, ln, pos) \quad (4)$ $teaches(iri2(icode), iri3(course)) \Leftarrow teaching(course, icode) \quad (5)$ FK: $\exists y_1. teaching(y_1, x) \rightarrow \exists y_2 y_3 y_4. academic(x, y_2, y_3, y_4)$

TBox, user-defined mapping assertions, and foreign key

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$$AssociateProfessor \sqsubseteq Person \quad AssociateProfessor(\mathbf{iri2}(acode)) \Leftarrow academic(acode, fn, ln, pos), pos = 2 \quad (3)$$

$$\exists teaches \sqsubseteq Person \quad FacultyMember(\mathbf{iri2}(acode)) \Leftarrow academic(acode, fn, ln, pos) \quad (4)$$

$$teaches(\mathbf{iri2}(acode), \mathbf{iri3}(course)) \Leftarrow teaching(course, acode) \quad (5)$$

$$FK: \exists y_1. teaching(y_1, x) \rightarrow \exists y_2 y_3 y_4. academic(x, y_2, y_3, y_4)$$

By **saturating the mapping** , we obtain mapping assertions for *Person*

$$Person(\mathbf{iri1}(scode)) \Leftarrow student(scode, fn, ln) \quad (6)$$

$$Person(\mathbf{iri2}(acode)) \Leftarrow academic(acode, fn, ln, pos), pos = 9 \quad (7)$$

$$Person(\mathbf{iri2}(acode)) \Leftarrow academic(acode, fn, ln, pos), pos = 2 \quad (8)$$

$$Person(\mathbf{iri2}(acode)) \Leftarrow academic(acode, fn, ln, pos) \quad (9)$$

$$Person(\mathbf{iri2}(acode)) \Leftarrow teaching(course, acode) \quad (10)$$

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By **optimizing the mapping** using query containment and the FK, we can remove mapping assertions 7, 8, and 10

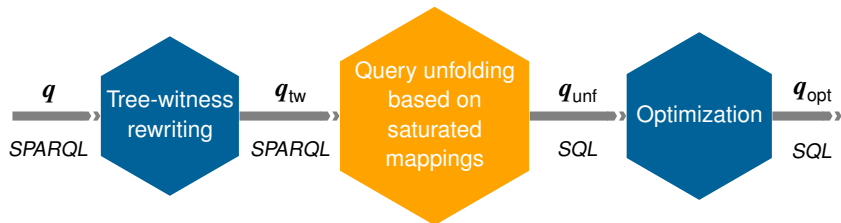
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Query reformulation as implemented by Ontop



| Step | Input | Output |
|---------------------------|--------------------------------------|-------------------|
| 1. Tree-witness rewriting | q (SPARQL) and \mathcal{T} | q_{tw} (SPARQL) |
| 2. Query unfolding | q_{tw} and \mathcal{M}_{sat} | q_{unf} (SQL) |
| 3. Query optimization | q_{unf} , primary and foreign keys | q_{opt} (SQL) |

SQL query optimization

Objective : produce SQL queries that are ...

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- adapted to existing query planners

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Semantic optimization

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

Integrity constraints

- Primary and foreign keys, unique constraints
- Sometimes implicit
- **Vital for query reformulation!**

Reformulation example – 1. Unfolding

Saturated mapping

$Teacher(iri2(acode)) \Leftarrow \text{academic}(acode, fn, ln, pos),$
 $pos \in [1..8]$

$Teacher(iri2(acode)) \Leftarrow \text{teaching}(course, acode)$

$firstName(iri1(scode), fn) \Leftarrow \text{student}(scode, fn, ln)$

$firstName(iri2(acode), fn) \Leftarrow \text{academic}(acode, fn, ln, pos)$

$lastName(iri1(scode), ln) \Leftarrow \text{student}(scode, fn, ln)$

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Query (we assume that the ontology is empty, hence
 $q_{tw} = q$)

$q(x, y, z) \leftarrow Teacher(x), firstName(x, y), lastName(x, z)$

Reformulation example – 1. Unfolding

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$$\text{lastName}(\text{iri2}(\text{acode}), \text{ln}) \Leftarrow \text{academic}(\text{acode}, \text{fn}, \text{ln}, \text{pos})$$

Query (we assume that the ontology is empty, hence

$$q_{\text{tw}} = q)$$

$$q(x, y, z) \Leftarrow \text{Teacher}(x), \text{firstName}(x, y), \text{lastName}(x, z)$$

Query unfolding

$$q_{\text{unf}}(x, y, z) \Leftarrow q^1_{\text{unf}}(x), q^2_{\text{unf}}(x, y),$$

$$q^3_{\text{unf}}(x, z)$$

$$q^1_{\text{unf}}(\text{iri2}(\text{acode})) \Leftarrow \text{academic}(\text{acode}, \text{fn}, \text{ln}, \text{pos}),$$

$$\text{pos} \in [1..8]$$

$$q^1_{\text{unf}}(\text{iri2}(\text{acode})) \Leftarrow \text{teaching}(\text{course}, \text{acode})$$

$$q^2_{\text{unf}}(\text{iri1}(\text{scode}), \text{fn}) \Leftarrow \text{student}(\text{scode}, \text{fn}, \text{ln})$$

$$q^2_{\text{unf}}(\text{iri2}(\text{acode}), \text{fn}) \Leftarrow \text{academic}(\text{acode}, \text{fn}, \text{ln}, \text{pos})$$

$$q^3_{\text{unf}}(\text{iri1}(\text{scode}), \text{ln}) \Leftarrow \text{student}(\text{scode}, \text{fn}, \text{ln})$$

$$q^3_{\text{unf}}(\text{iri2}(\text{acode}), \text{ln}) \Leftarrow \text{academic}(\text{acode}, \text{fn}, \text{ln}, \text{pos})$$

Reformulation example – 1. Unfolding

Saturated mapping

$Teacher(iri2(acode)) \Leftarrow academic(acode, fn, ln, pos),$
 $pos \in [1..8]$

$Teacher(iri2(acode)) \Leftarrow teaching(course, acode)$

$firstName(iri1(scode), fn) \Leftarrow student(scode, fn, ln)$

$firstName(iri2(acode), fn) \Leftarrow academic(acode, fn, ln, pos)$

$lastName(iri1(scode), ln) \Leftarrow student(scode, fn, ln)$

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Query (we assume that the ontology is empty, hence

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$q(x, y, z) \Leftarrow Teacher(x), firstName(x, y), lastName(x, z)$

Query unfolding, and **normalization**, to make the join conditions explicit

$q_{norm}(x, y, z) \Leftarrow q^1_{unf}(x), q^2_{unf}(x_1, y),$
 $q^3_{unf}(x_2, z), x = x_1, x = x_2$

$q^1_{unf}(iri2(acode)) \Leftarrow academic(acode, fn, ln, pos),$
 $pos \in [1..8]$

$q^1_{unf}(iri2(acode)) \Leftarrow teaching(course, acode)$

$q^2_{unf}(iri1(scode), fn) \Leftarrow student(scode, fn, ln)$

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$q^3_{unf}(iri2(acode), ln) \Leftarrow academic(acode, fn, ln, pos)$

Reformulation example – 2. Structural optimization

Unfolded normalized query

$$\mathbf{q}_{\text{norm}}(x, y, z) \leftarrow \mathbf{q}^1_{\text{unf}}(x), \mathbf{q}^2_{\text{unf}}(x_1, y), \\ \mathbf{q}^3_{\text{unf}}(x_2, z), \\ x = x_1, x = x_2$$
$$\mathbf{q}^1_{\text{unf}}(\mathbf{iri2}(a)) \leftarrow \text{academic}(a, f, l, p), \\ p \in [1..8]$$
$$\mathbf{q}^1_{\text{unf}}(\mathbf{iri2}(a)) \leftarrow \text{teaching}(c, a)$$
$$\mathbf{q}^2_{\text{unf}}(\mathbf{iri1}(s), f) \leftarrow \text{student}(s, f, l)$$
$$\mathbf{q}^2_{\text{unf}}(\mathbf{iri2}(a), f) \leftarrow \text{academic}(a, f, l, p)$$
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Reformulation example – 2. Structural optimization

Unfolded normalized query

$$q_{\text{norm}}(x, y, z) \leftarrow q^1_{\text{unf}}(x), q^2_{\text{unf}}(x_1, y), \\ q^3_{\text{unf}}(x_2, z), \\ x = x_1, x = x_2$$

$$q^1_{\text{unf}}(\text{iri2}(a)) \leftarrow \text{academic}(a, f, l, p), \\ p \in [1..8]$$

$$q^1_{\text{unf}}(\text{iri2}(a)) \leftarrow \text{teaching}(c, a)$$

$$q^2_{\text{unf}}(\text{iri1}(s), f) \leftarrow \text{student}(s, f, l)$$

$$q^2_{\text{unf}}(\text{iri2}(a), f) \leftarrow \text{academic}(a, f, l, p)$$

$$q^3_{\text{unf}}(\text{iri1}(s), l) \leftarrow \text{student}(s, f, l)$$

$$q^3_{\text{unf}}(\text{iri2}(a), l) \leftarrow \text{academic}(a, f, l, p)$$

Flattening (URI template lifting) – Part 1/2

$$q_{\text{lift}}(\text{iri2}(a), y, z) \leftarrow \text{academic}(a, f_1, l_1, p_1), \\ \text{student}(s, f_2, l_2), \\ \text{student}(s_1, f_3, l_3), \\ \text{iri2}(a) = \text{iri1}(s), \\ \text{iri2}(a) = \text{iri1}(s_1), \\ p_1 \in [1..8]$$

$$q_{\text{lift}}(\text{iri2}(a), y, z) \leftarrow \text{academic}(a, f_1, l_1, p_1), \\ \text{student}(s, f_2, l_2), \\ \text{academic}(a_2, f_3, z, p_3), \\ \text{iri2}(a) = \text{iri1}(s), \\ \text{iri2}(a) = \text{iri2}(a_2), \\ p_1 \in [1..8]$$

(One sub-query not shown)

$$q_{\text{lift}}(\text{iri2}(a), y, z) \leftarrow \text{academic}(a, f_1, l_1, p_1), \\ \text{academic}(a_1, y, l_2, p_2), \\ \text{academic}(a_2, f_3, z, p_3), \\ \text{iri2}(a) = \text{iri2}(a_1), \\ \text{iri2}(a) = \text{iri2}(a_2), \\ p_1 \in [1..8]$$

Reformulation example – 2. Structural optimization

Unfolded normalized query

$$q_{\text{norm}}(x, y, z) \leftarrow q^1_{\text{unf}}(x), q^2_{\text{unf}}(x_1, y), \\ q^3_{\text{unf}}(x_2, z), \\ x = x_1, x = x_2$$

$$q^1_{\text{unf}}(\text{iri2}(a)) \leftarrow \text{academic}(a, f, l, p), \\ p \in [1..8]$$

$$q^1_{\text{unf}}(\text{iri2}(a)) \leftarrow \text{teaching}(c, a)$$

$$q^2_{\text{unf}}(\text{iri1}(s), f) \leftarrow \text{student}(s, f, l)$$

$$q^2_{\text{unf}}(\text{iri2}(a), f) \leftarrow \text{academic}(a, f, l, p)$$

$$q^3_{\text{unf}}(\text{iri1}(s), l) \leftarrow \text{student}(s, f, l)$$

$$q^3_{\text{unf}}(\text{iri2}(a), l) \leftarrow \text{academic}(a, f, l, p)$$

- While flattening, we can avoid to generate those queries that contain in their body an equality between two terms with incompatible IRI templates.
- This might avoid a potential exponential blowup.

Flattening (URI template lifting) – Part 2/2

$$q^{\text{lift}}_{\text{iri2}}(\text{iri2}(a), y, z) \leftarrow \text{teaching}(c, a), \\ \text{student}(s, f_2, l_2), \\ \text{student}(s_1, f_3, l_3), \\ \text{iri2}(a) = \text{iri1}(s), \\ \text{iri2}(a) = \text{iri1}(s_1)$$

$$q^{\text{lift}}_{\text{iri2}}(\text{iri2}(a), y, z) \leftarrow \text{teaching}(c, a), \\ \text{student}(s, f_2, l_2), \\ \text{academic}(a_2, f_3, z, p_3), \\ \text{iri2}(a) = \text{iri1}(s), \\ \text{iri2}(a) = \text{iri2}(a_2)$$

(One sub-query not shown)

$$q^{\text{lift}}_{\text{iri2}}(\text{iri2}(a), y, z) \leftarrow \text{teaching}(c, a), \\ \text{academic}(a_1, y, l_2, p_2), \\ \text{academic}(a_2, f_3, z, p_3), \\ \text{iri2}(a) = \text{iri2}(a_1), \\ \text{iri2}(a) = \text{iri2}(a_2)$$

Reformulation example – 3. Semantic optimization

We are left with just two queries

$q_{\text{lift}}(\text{iri2}(a), y, z) \leftarrow \text{academic}(a, f_1, l_1, p_1), p_1 \in [1..8],$
 $\text{academic}(a_1, y, l_2, p_2), \text{iri2}(a) = \text{iri2}(a_1),$
 $\text{academic}(a_2, f_3, z, p_3), \text{iri2}(a) = \text{iri2}(a_2)$

$q_{\text{lift}}(\text{iri2}(a), y, z) \leftarrow \text{teaching}(c, a),$
 $\text{academic}(a_1, y, l_2, p_2), \text{iri2}(a) = \text{iri2}(a_1),$
 $\text{academic}(a_2, f_3, z, p_3), \text{iri2}(a) = \text{iri2}(a_2)$

Reformulation example – 3. Semantic optimization

We are left with just two queries, that we can simplify by eliminating equalities

$q_{\text{struct}}(\text{iri2}(a), y, z) \leftarrow \text{academic}(a, f_1, l_1, p_1), p_1 \in [1..8],$
 $\text{academic}(a, y, l_2, p_2),$
 $\text{academic}(a, f_3, z, p_3)$

$q_{\text{struct}}(\text{iri2}(a), y, z) \leftarrow \text{teaching}(c, a),$
 $\text{academic}(a, y, l_2, p_2),$
 $\text{academic}(a, f_3, z, p_3)$

Reformulation example – 3. Semantic optimization

We are left with just two queries, that we can simplify by eliminating equalities

$$q_{\text{struct}}(\text{iri2}(a), y, z) \leftarrow \text{academic}(a, f_1, l_1, p_1), p_1 \in [1..8],$$
$$\text{academic}(a, y, l_2, p_2),$$
$$\text{academic}(a, f_3, z, p_3)$$
$$q_{\text{struct}}(\text{iri2}(a), y, z) \leftarrow \text{teaching}(c, a),$$
$$\text{academic}(a, y, l_2, p_2),$$
$$\text{academic}(a, f_3, z, p_3)$$

We can then exploit database constraints (e.g., primary keys) for semantic optimization of the query.

Self-join elimination (semantic optimization)

PK: $\text{academic}(a, f, l, p) \wedge \text{academic}(a, f', l', p')$
 $\rightarrow (f = f') \wedge (l = l') \wedge (p = p')$

Reformulation example – 3. Semantic optimization

We are left with just two queries, that we can simplify by eliminating equalities

$$q_{\text{struct}}(\text{iri2}(a), y, z) \leftarrow \text{academic}(a, f_1, l_1, p_1), p_1 \in [1..8],$$

$$\text{academic}(a, y, l_2, p_2),$$

$$\text{academic}(a, f_3, z, p_3)$$

$$q_{\text{struct}}(\text{iri2}(a), y, z) \leftarrow \text{teaching}(c, a),$$

$$\text{academic}(a, y, l_2, p_2),$$

$$\text{academic}(a, f_3, z, p_3)$$

We can then exploit database constraints (e.g., primary keys) for semantic optimization of the query.

Self-join elimination (semantic optimization)

PK: $\text{academic}(a, f, l, p) \wedge \text{academic}(a, f', l', p')$
 $\rightarrow (f = f') \wedge (l = l') \wedge (p = p')$

$$q_{\text{opt}}(\text{iri2}(a), y, z) \leftarrow \text{academic}(a, y, z, p_1), p_1 \in [1..8]$$

$$q_{\text{opt}}(\text{iri2}(a), y, z) \leftarrow \text{teaching}(c, a), \text{academic}(a, y, z, p_2)$$

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