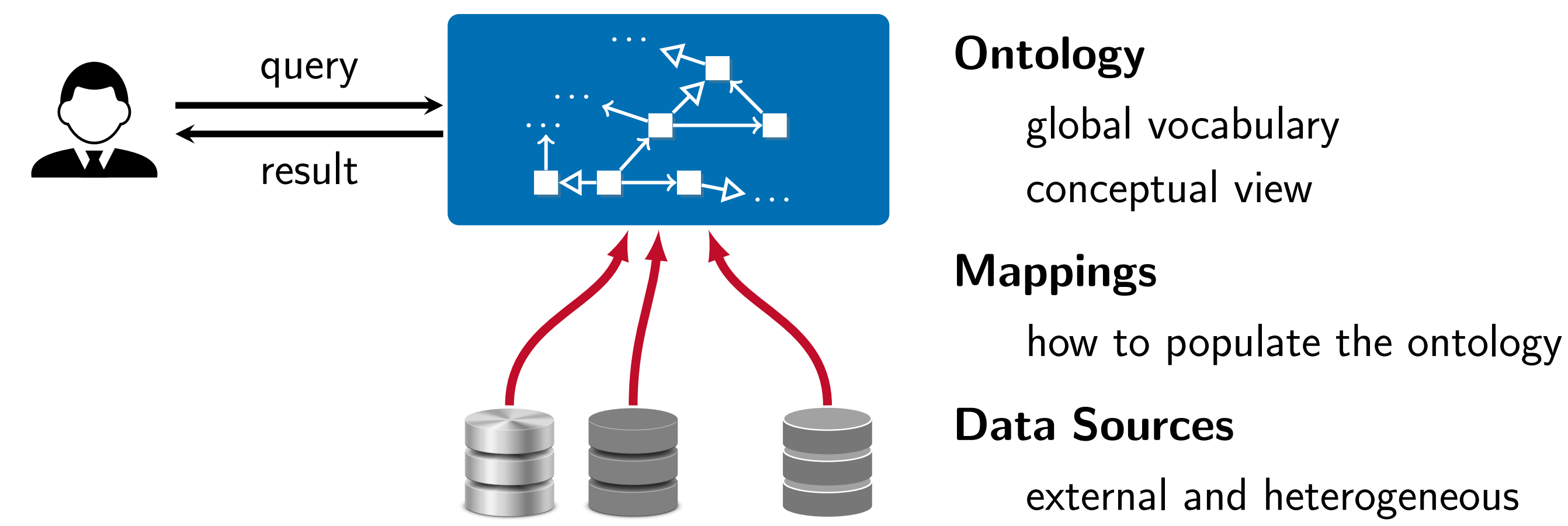


Ontology Based Data Access – OBDA



Logical transparency in accessing data:

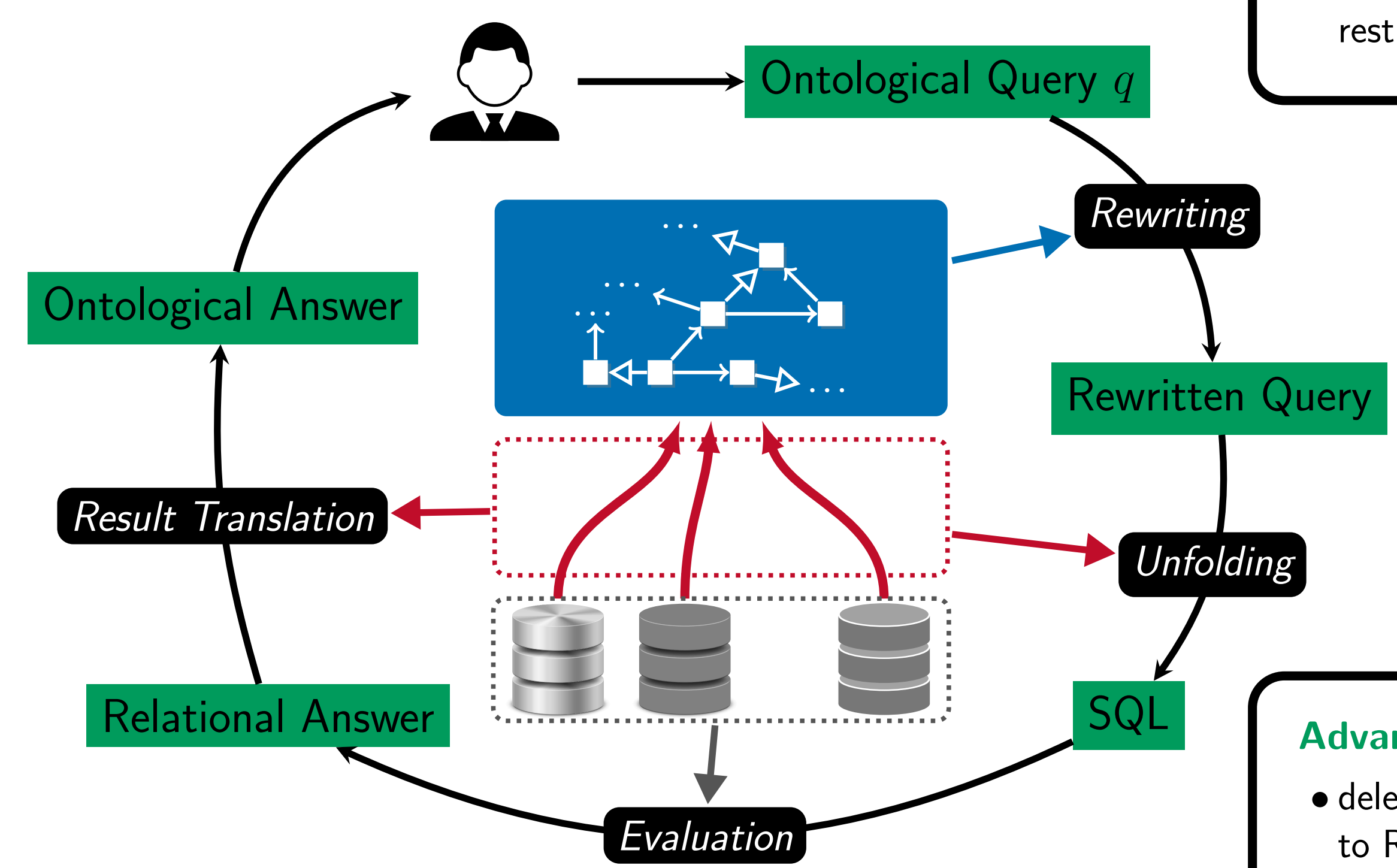
- does not know where and how data is stored;
- can only see a **conceptual view** of data.

An **OBDA specification** is a triple $\langle \mathcal{T}, \mathcal{M}, \mathcal{S} \rangle$, where

- \mathcal{T} is a Description Logic TBox,
- \mathcal{M} is a set of mapping assertions $SQL_A(x) \rightsquigarrow A(x)$ and $SQL_R(x, y) \rightsquigarrow R(x, y)$, and
- \mathcal{S} is a relational schema.

Virtual Approach to OBDA

Query Answering is done by **Query Rewriting** into SQL.



Disadvantages:

- works only for first-order (FO) rewritable languages: the commonly adopted one is OWL 2 QL, which is too restrictive

Advantages:

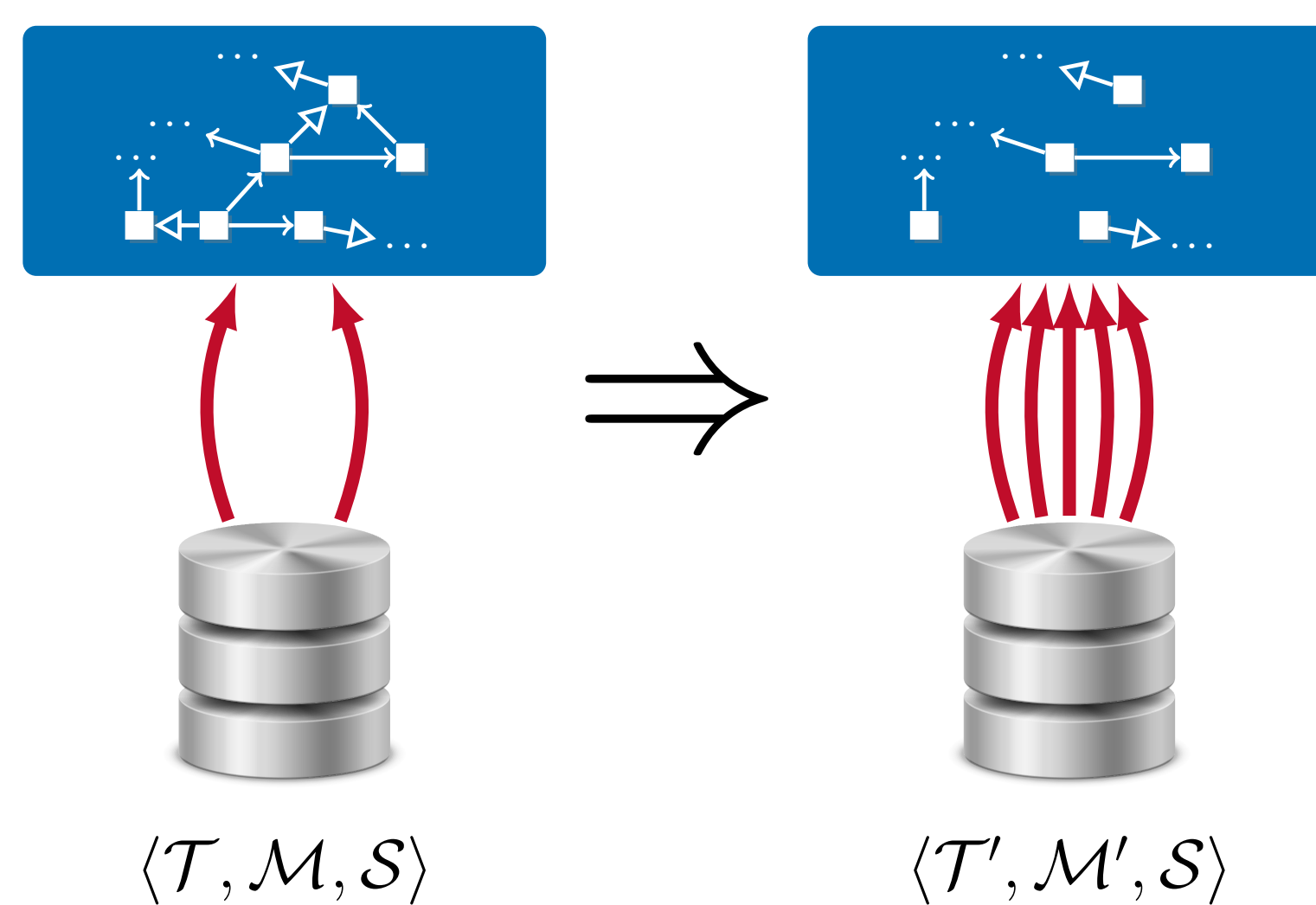
- delegates query evaluation to RDBMS
- avoids data materialization
- works well for Big Data

Beyond OWL 2 QL: Mappings to the Rescue

Problem We want to go beyond OWL 2 QL in Virtual OBDA. However expressive ontologies are **not** FO-rewritable.

Solution Exploit the **mapping component** that makes use of arbitrary SQL queries.

We introduce a **framework** for rewriting and approximation of OBDA specifications.



$$\begin{aligned} \mathcal{T} &= \{ A \sqcap B \sqsubseteq C \} \\ \mathcal{M} &= \{ SQL_A(x) \rightsquigarrow A(x), \\ &\quad SQL_B(x) \rightsquigarrow B(x) \} \end{aligned} \Rightarrow \begin{aligned} \mathcal{T}' &= \{ \} \\ \mathcal{M}' &= \{ SQL_A(x) \rightsquigarrow A(x), \\ &\quad SQL_B(x) \rightsquigarrow B(x), \\ &\quad SQL_A(x) \wedge SQL_B(x) \rightsquigarrow C(x) \} \end{aligned}$$

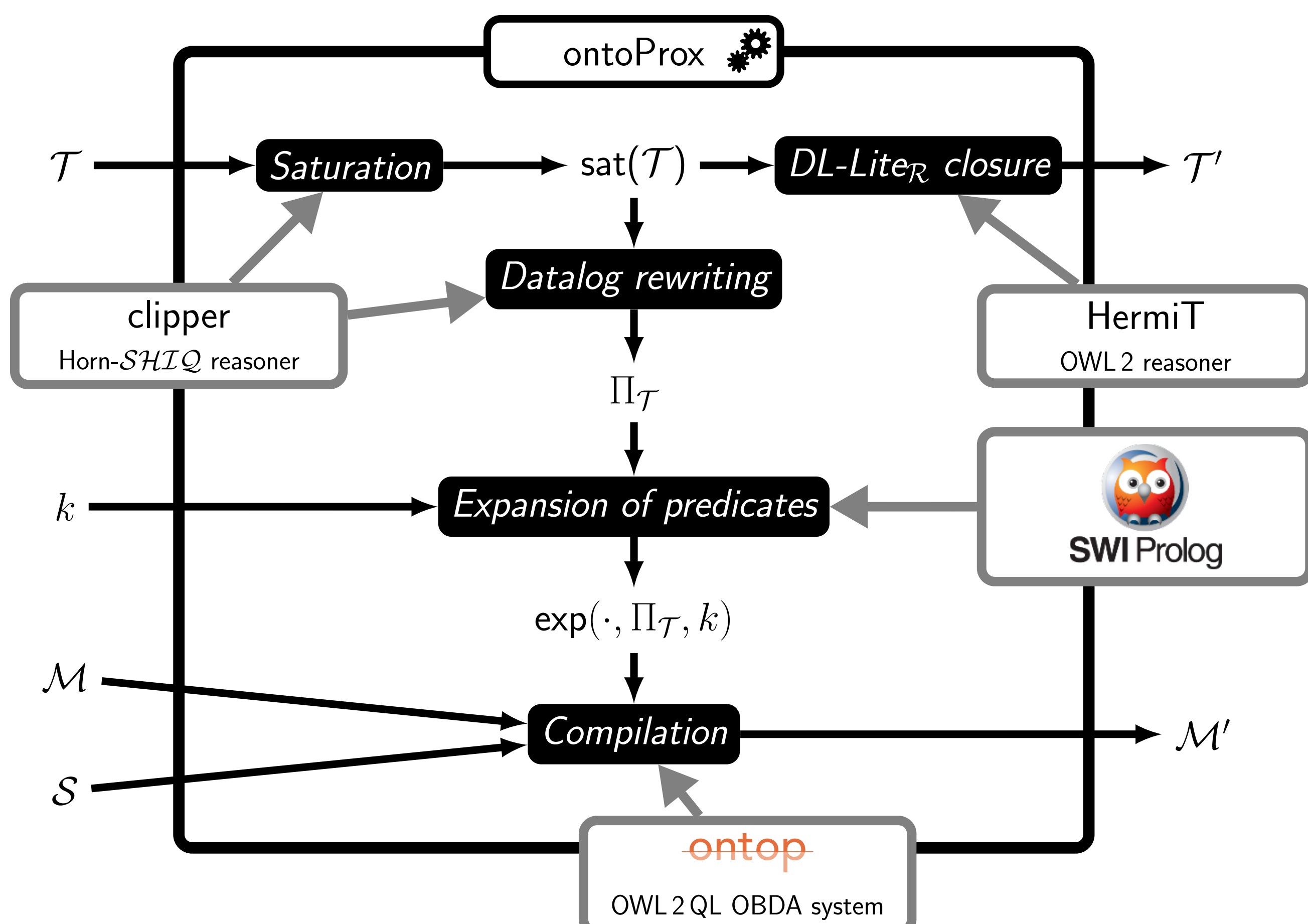
$$\begin{aligned} \mathcal{T} &= \{ \exists R.A \sqsubseteq C \} \\ \mathcal{M} &= \{ SQL_A(x) \rightsquigarrow A(x), \\ &\quad SQL_R(x, y) \rightsquigarrow R(x, y) \} \end{aligned} \Rightarrow \begin{aligned} \mathcal{T}' &= \{ \} \\ \mathcal{M}' &= \{ SQL_A(x) \rightsquigarrow A(x), \\ &\quad SQL_R(x, y) \rightsquigarrow R(x, y), \\ &\quad SQL_R(x, y) \wedge SQL_A(y) \rightsquigarrow C(x) \} \end{aligned}$$

$$\begin{aligned} \mathcal{T} &= \{ \exists R.A \sqsubseteq A \} \\ \mathcal{M} &= \{ SQL_A(x) \rightsquigarrow A(x), \\ &\quad SQL_R(x, y) \rightsquigarrow R(x, y) \} \end{aligned} \Rightarrow$$

$$\begin{aligned} \mathcal{T}' &= \{ \} \\ \mathcal{M}' &= \{ SQL_A(x) \rightsquigarrow A(x), \\ &\quad SQL_R(x, y) \rightsquigarrow R(x, y), \\ &\quad SQL_R(x, y) \wedge SQL_A(y) \rightsquigarrow A(x) \\ &\quad SQL_R(x, y) \wedge SQL_R(y, z) \wedge SQL_A(z) \rightsquigarrow A(x) \\ &\quad SQL_R(x, y) \wedge SQL_R(y, z) \wedge SQL_R(z, w) \wedge SQL_A(w) \rightsquigarrow A(x) \} \end{aligned}$$

ontoProx

We have developed an algorithm for computing approximations and implemented it in a prototype system called **ontoProx**.



It takes as input a Horn-SHIQ TBox \mathcal{T} , a mapping \mathcal{M} and an integer k , and produces a $DL-Lite_R$ TBox \mathcal{T}' and an extended mapping \mathcal{M}' .

Evaluation

We evaluated **ontoProx** over synthetic and real OBDA instances against

- the default **ontop** behavior,
- local semantic approximation (LSA),
- global semantic approximation (GSA), and
- **clipper** over materialized ABoxes.

The evaluation showed that we are able to obtain more answers using our approach (in fact, complete, whenever that could be verified by clipper).

<http://ontop.inf.unibz.it>

<https://github.com/ontop/ontoprox>

<https://github.com/ghxiao/clipper>