Graph-based Virtual Data Integration

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Data Integration

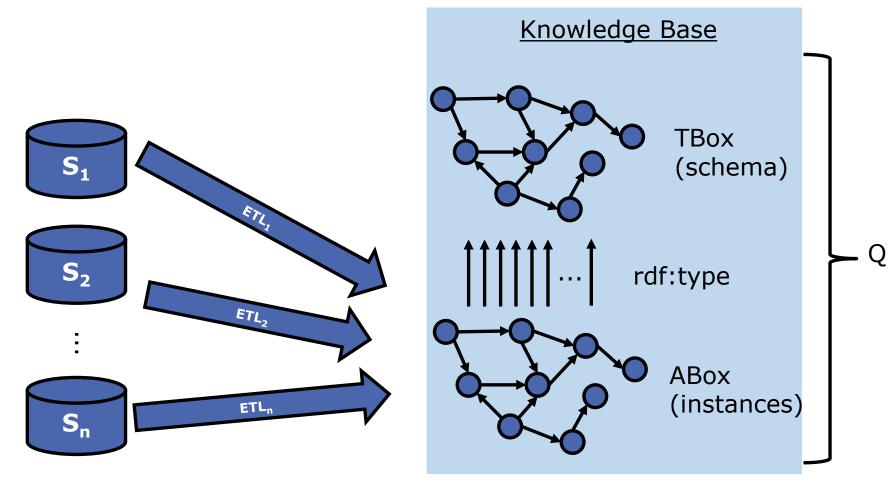
Data integration is an area of study within data management aimed at facilitating transparent access to a variety of heterogeneous data sources

- Two main options to perform data integration:
 - Physical data integration
 - Virtual data integration

In this course we will focus on using graphs to solve data integration

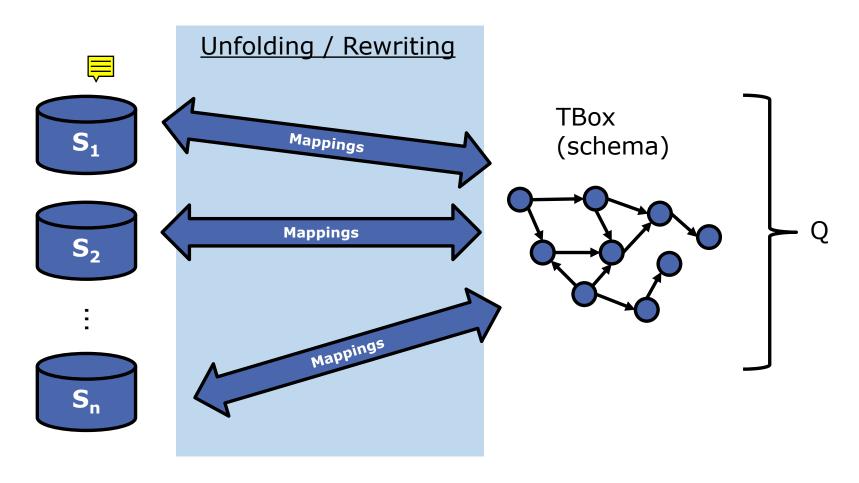
Graph-Based Data Integration at a Glance

Option 1: Physical Data Integration



Graph-Based Data Integration at a Glance

Option 2: Virtual Data Integration



Why Virtual Data Integration?

- When the data sources are not under your control and owners require a federation (e.g., data exchange between companies)
 - E.g., Data Portability
- When we do not want to move the data from where it resides
 - For example, key-based models are more performant than graph models for table scans
- When data freshness is crucial
 - ETLs run from time to time and the period between updates is called the update window
- Virtual data integration is simpler to maintain (most of the work resides on the rewriting algorithm)

Why Virtual Data Integration?

- Virtual Integration is very trendy in Data Science because we can use graph-based models to perform rich and complex data integration while benefitting from sequential reads when querying the data (which can reside in the most appropriate data storage)
- Many Big Data Integration systems work under this assumption
 - Data Tamer (<u>https://www.tamr.com/</u>)
 - The BigDAWG Polystore System (https://dl.acm.org/citation.cfm?id=3226620)
 - Ontop (http://ontop.inf.unibz.it/)
 - ODIN (http://www.essi.upc.edu/~snadal/odin.html)

Two Main Approaches

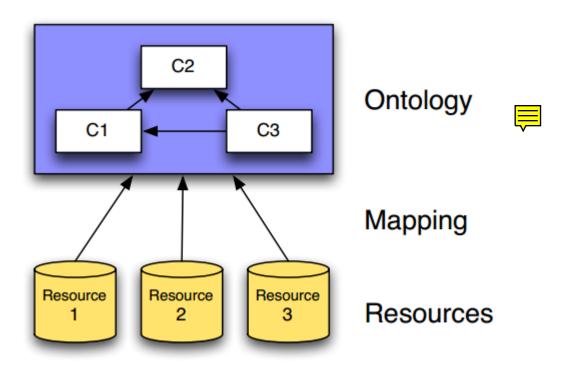
- Ontology-based data access
 - Monolithic approach
 - The TBox is directly related to the sources via mappings
- Ontology-mediated queries
 - Relies on the concept of wrapper
 - Thus, we can select a subset of the data source to be exposed to the whole integration System
 - Security
 - Modularity
 - It allows pay-as-you-go data Integration
 - The integrated schema is built incrementally as new data sources arrive

GAV Data Integration

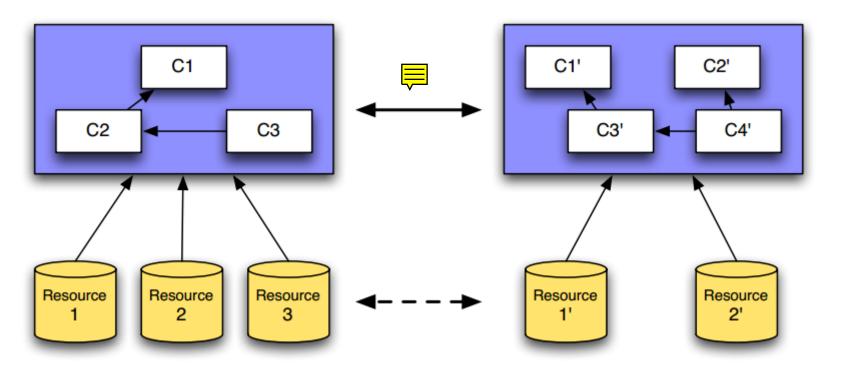
ONTOLOGY-BASED DATA ACCESS

Ontology-Based Data Access

Ontology-mediated data access



An Approach for Data Integration



The DL-Lite Family



- Right trade-off between expressiveness and data complexity query answering
 - PTime in the size of the TBOX
 - LogSpace in the size of the ABOX
- Two maximal DLs satisfying this trade-off
 - DL-Lite_F
 - DL-Lite_R



Remember the DL-Lite family maps to **OWL 2 QL**

DL-Lite_F

TBox assertions:

• Concept inclusion assertions: $Cl \subseteq Cr$, with:

• Functionality assertions: (funct Q)

ABox assertions: A(c), $P(c_1, c_2)$, with c_1 , c_2 constants

Observations:

- Captures all the basic constructs of UML Class Diagrams and ER
- Notable exception: covering constraints in generalizations.

Semantics of DL-Lite

■ It basically captures the expressivity of a UML class diagram

ISA between classes	A_1 !	$\sqsubseteq A_2$
Disjointness between classes	$A_1 \sqsubseteq \neg A_2$	
Domain and range of relations	$\exists P \sqsubseteq A_1$	$\exists P^- \sqsubseteq A_2$
Mandatory participation	$A_1 \sqsubseteq \exists P$	$A_2 \sqsubseteq \exists P^-$
Functionality of relations (in DL - $Lite_{\mathcal{F}}$)	(funct $P)$	$(\operatorname{funct} P^-)$
ISA between relations (in DL - $Lite_{\mathcal{R}}$)	$Q_1 \sqsubseteq Q_2$	
Disjointness between relations (in DL - $Lite_{\mathcal{R}}$)	$Q \sqsubseteq \neg Q$	

Construct	Syntax	Example	Semantics
atomic conc.	A	Doctor	$A^{\mathcal{I}} \subseteq \Delta^{\mathcal{I}}$
exist. restr.	$\exists Q$	∃child [—]	$\{d \mid \exists e. (d, e) \in Q^{\mathcal{I}}\}$
at. conc. neg.	$\neg A$	$\neg Doctor$	$\Delta^{\mathcal{I}} \setminus A^{\mathcal{I}}$
conc. neg.	$\neg \exists Q$	¬∃child	$\Delta^{\mathcal{I}} \setminus (\exists Q)^{\mathcal{I}}$
atomic role	P	child	$P^{\mathcal{I}} \subseteq \Delta^{\mathcal{I}} \times \Delta^{\mathcal{I}}$
inverse role	P^-	child ⁻	$\{(o,o')\mid (o',o)\in P^{\mathcal{I}}\}$
role negation	$\neg Q$	¬manages	$(\Delta_O^{\mathcal{I}} \times \Delta_O^{\mathcal{I}}) \setminus Q^{\mathcal{I}}$
conc. incl.	$Cl \sqsubseteq Cr$	Father ⊑ ∃child	$Cl^{\mathcal{I}} \subseteq Cr^{\mathcal{I}}$
role incl.	$Q \sqsubseteq R$	$hasFather \sqsubseteq child$	$Q^{\mathcal{I}} \subseteq R^{\mathcal{I}}$
funct. asser.	$(\mathbf{funct}\ Q)$	(funct succ)	$\forall d, e, e'. (d, e) \in Q^{\mathcal{I}} \land (d, e') \in Q^{\mathcal{I}} \rightarrow e = e'$
mem. asser.	A(c)	Father(bob)	$c^{\mathcal{I}} \in A^{\mathcal{I}}$
mem. asser.	$P(c_1, c_2)$	child(bob, ann)	$(c_1^{\mathcal{I}}, c_2^{\mathcal{I}}) \in P^{\mathcal{I}}$

Linking Data to Ontologies

- The ABOX is stored in a relational database
 - OBDA has recently been extended to other kind of sources, like document-stores
- Direct mappings between the TBOX and DB
 - Query answering is reformulated in terms of the TBOX, a set of mappings and a RDBMS

Theorem

Query answering in a DL-Lite $_A$ OBDM system $\mathcal{O} = \langle \mathcal{T}, \mathcal{M}, \mathcal{D} \rangle$ is

- NP-complete in the size of the query.
- **2** PTIME in the size of the TBox \mathcal{T} and the mappings \mathcal{M} .
- **3** LogSpace in the size of the database \mathcal{D} .

Mappings

- OBDA works with GAV mappings
- Typically, they use RDF-based mapping languages to express them
 - R2RML (a language to express mappings from global concepts to relational databases)

```
mappingId Actor
target imdb:name/{person_id} a dbpedia:Actor .
source select person id from cast info where cast info.role id = 1
```

RML is a generalisation to map to any kind of source (http://rml.io/)

A Tool for OBDA

- Ontop: http://ontop.inf.unibz.it/
 - OWL 2 QL
 - RDFS



Code, examples and more:

https://github.com/ontop/ontop

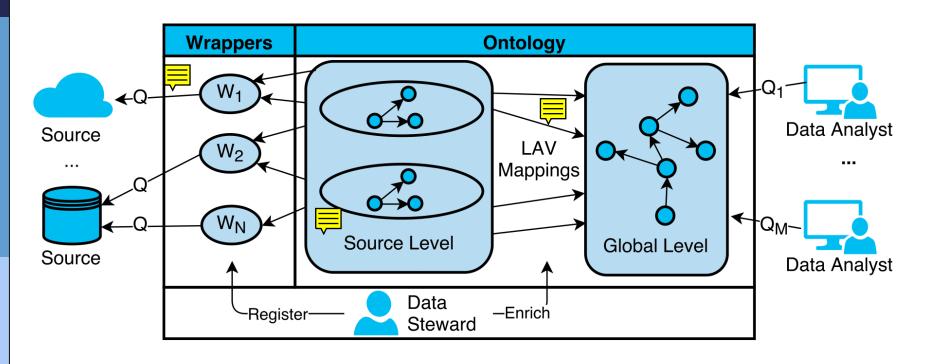
LAV Data Integration

ONTOLOGY-MEDIATED QUERIES

OMQ

- It is a family of systems performing graphbased data integration with LAV
 - Conceptually, GAV is also possible
- Based on the well-known wrappermediator architecture
- To make the querying rewriting feasible, they adopt several measures:
 - Exact mappings (i.e., Closed-World assumption)
 - Very basic reasoning capabilities (taxonomies and domain / range inference)

Ontology-mediated Query



Virtual integration with LAV mappings

Big Data Integration Ontology

- We revisit the Data Integration framework and construct an ontology as follows:
 - Global level (G) Integrated view
 - Source levels (S) Views on the data sources (wrappers)
 - Mappings (M) LAV mappings between G and S
- Example:
 - Cross-domain queries on:
 - Monitored data on video players (lag ratio, etc.)
 - Tweets in English gathered through a feedback gathering tool

Wrappers

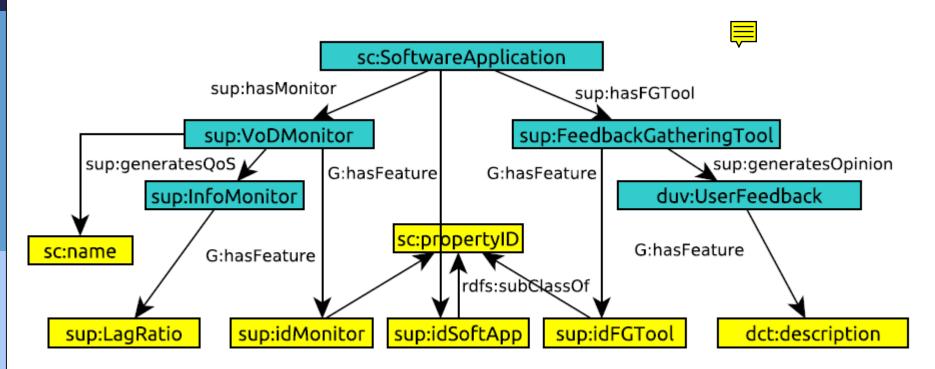
- They represent a view on the source
- You can think of a <u>named</u> query over the source. For example:

W1: SELECT a, b, c FROM T

- Typical assumptions made by wrappers:
 - They expose the source in **tabular format** (1NF)
 - Thus, Cypher, SPARQL or MongoDB's aggregation framework would also meet the requirements
- In general, most query languages produce tabular format
- A data source may generate several wrappers
- Typically, new versions of data are considered new wrappers

Global Level

- ☐ Green: concepts
- Yellow: attributes



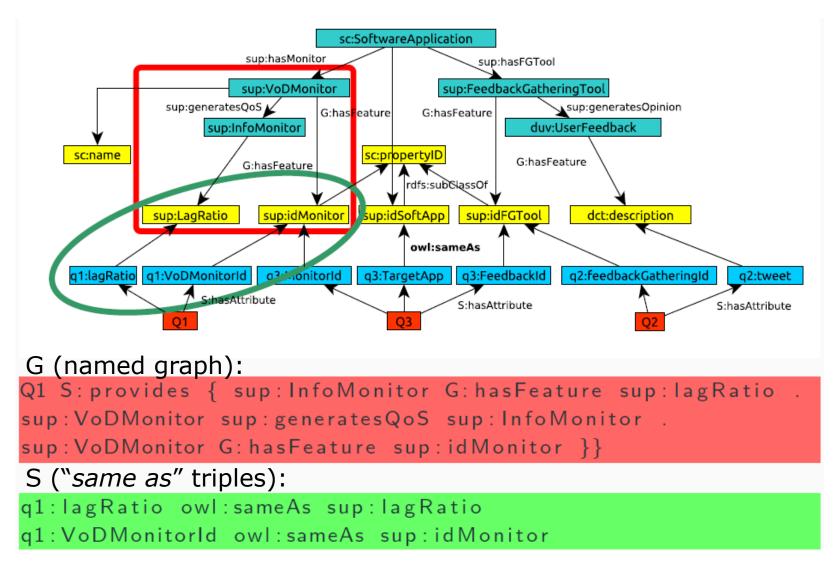
- Sources are exposed by means of wrappers
 - We automatically bootstrap the attributes projected by the wrappers

```
Q_1: \  \, \mathsf{ID} \  \, \mathsf{and} \  \, \mathsf{compute} \  \, \mathsf{the} \  \, \mathsf{lag} \  \, \mathsf{ratio} \\ \mathsf{db.} \  \, \mathsf{getCollection} \  \, ('\mathsf{vod}'). \  \, \mathsf{aggregate} \  \, ([\\ \{\$\mathsf{project:} \  \, \{"\mathsf{VoDmonitorId}":\mathsf{true} \  \, , "\mathsf{lagRatio}": \, \{\$\mathsf{divide} \  \, : \, ["\,\$\mathsf{waitTime}","\,\$\mathsf{watchTime}"]\}\}] \\ \mathsf{Q}_2 \  \, \mathsf{-all} \  \, \mathsf{attributes} \  \, \mathsf{for} \  \, \mathsf{tweets} \  \, \mathsf{in} \  \, \mathsf{english}. \\ \mathsf{Q}_3 \  \, \mathsf{-association} \  \, \mathsf{target} \  \, \mathsf{app} \  \, \to \  \, \mathsf{monitor}, \  \, \mathsf{feedback} \  \, \mathsf{gathering} \  \, \mathsf{tool} \\ \mathsf{q}_1: \mathsf{lagRatio} \  \, \mathsf{q}_1: \mathsf{lagRatio} \  \, \mathsf{q}_2: \mathsf{feedbackGatheringid} \  \, \mathsf{q}_2: \mathsf{feedbackGatheringid} \  \, \mathsf{q}_2: \mathsf{feedbackGatheringid} \  \, \mathsf{q}_2: \mathsf{feedbackGatheringid} \  \, \mathsf{q}_3: \mathsf{feedbackGatheringid} \  \, \mathsf{q}_
```

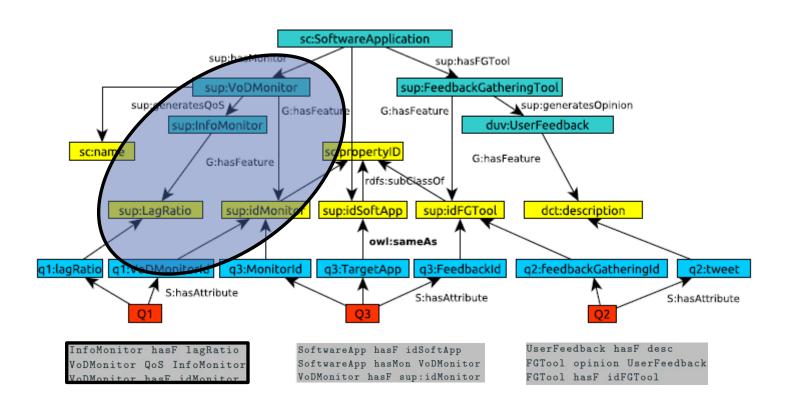
Red: Wrappers; Blue: Wrapper attributes

- A LAV mapping for a wrapper Q is defined as: $M = \langle G, S \rangle$ where:
 - \blacksquare G is a named graph \blacksquare
 - S is a set of triples of the form:
 - <x, owl:sameAs, y>, where
 - <x, rdf:type, S:Attribute> and
 - < < y, rdf:type, G:Feature>

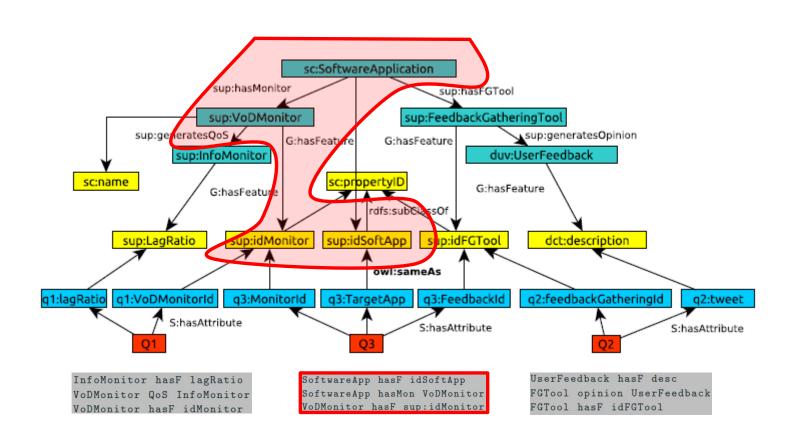
LAV Mapping Example



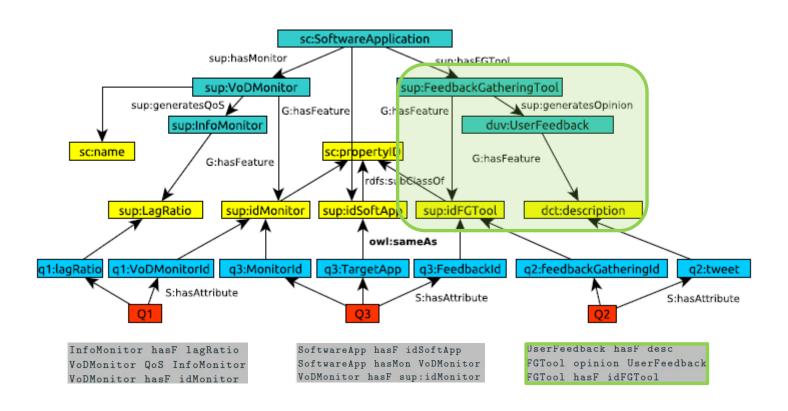
LAV Mappings (Q1)



LAV Mappings (Q3)



LAV Mappings (Q2)



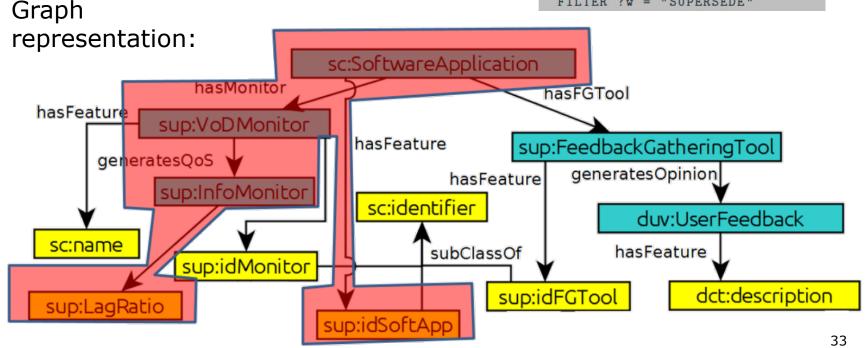
Query Answering – Rewriting Algorithm

- Any SPARQL query on the global graph must be rewritten as a query in terms of the wrappers
- Example of query over G:

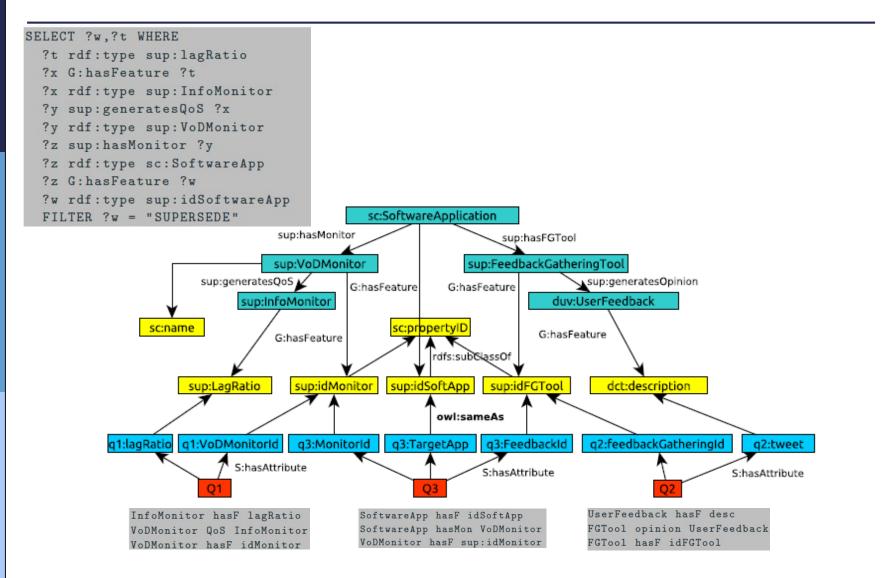
SPARQL Query:

```
SELECT ?w,?t WHERE

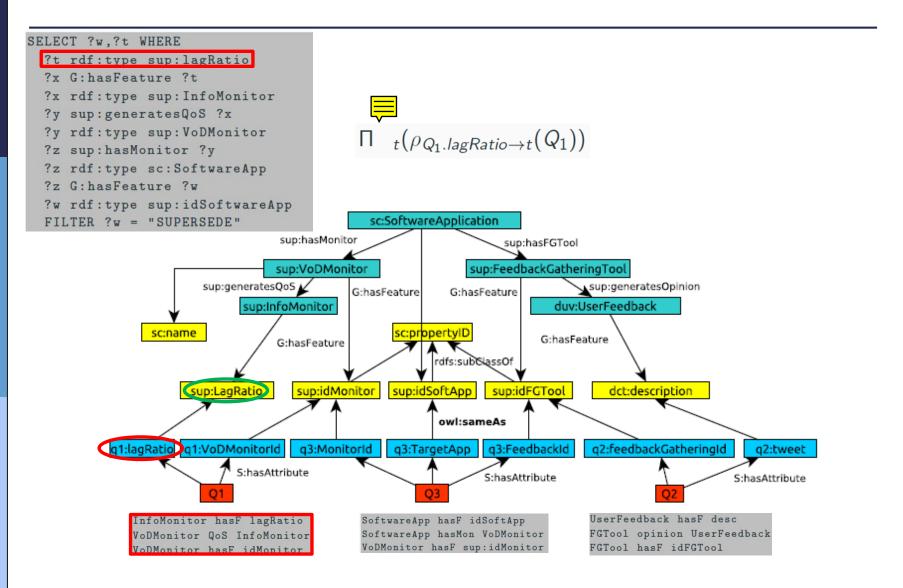
?t rdf:type sup:lagRatio
?x G:hasFeature ?t
?x rdf:type sup:InfoMonitor
?y sup:generatesQoS ?x
?y rdf:type sup:VoDMonitor
?z sup:hasMonitor ?y
?z rdf:type sc:SoftwareApp
?z G:hasFeature ?w
?w rdf:type sup:idSoftwareApp
FILTER ?w = "SUPERSEDE"
```



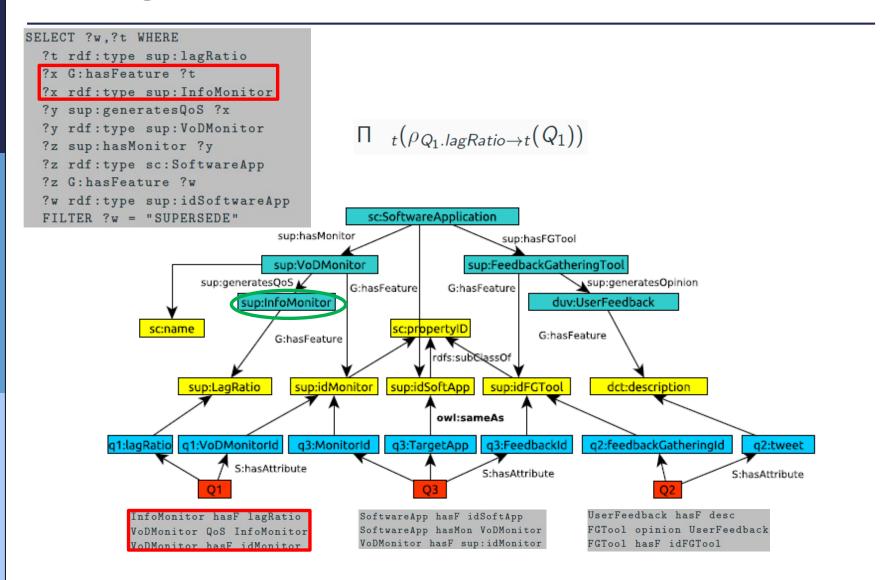
Notions on the Query Rewriting Alg.



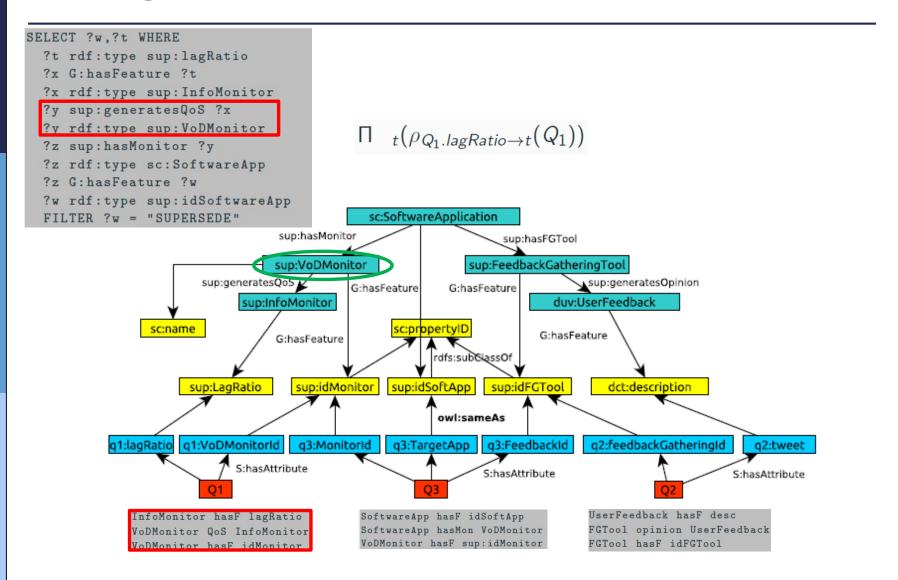
Start from a Terminal Feature



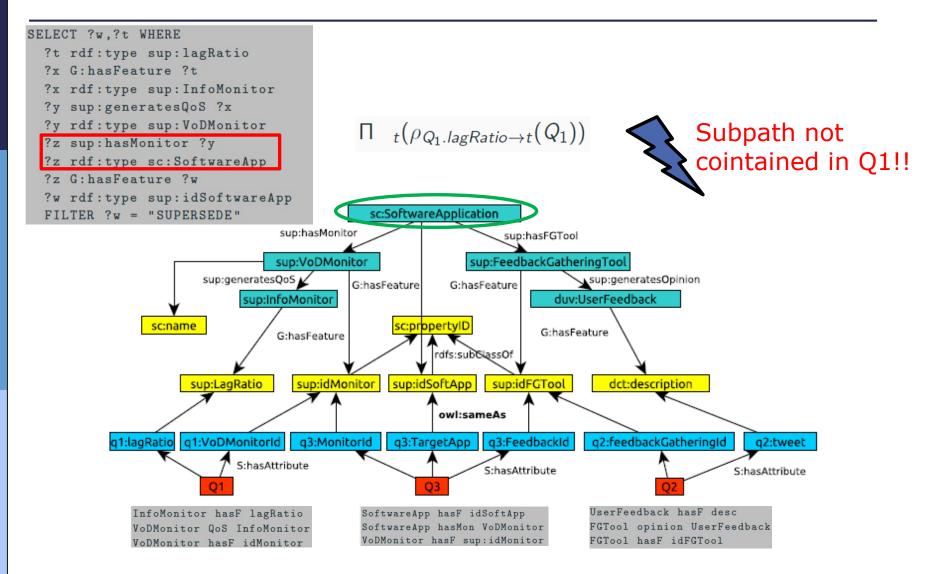
Navigate G from the Feature



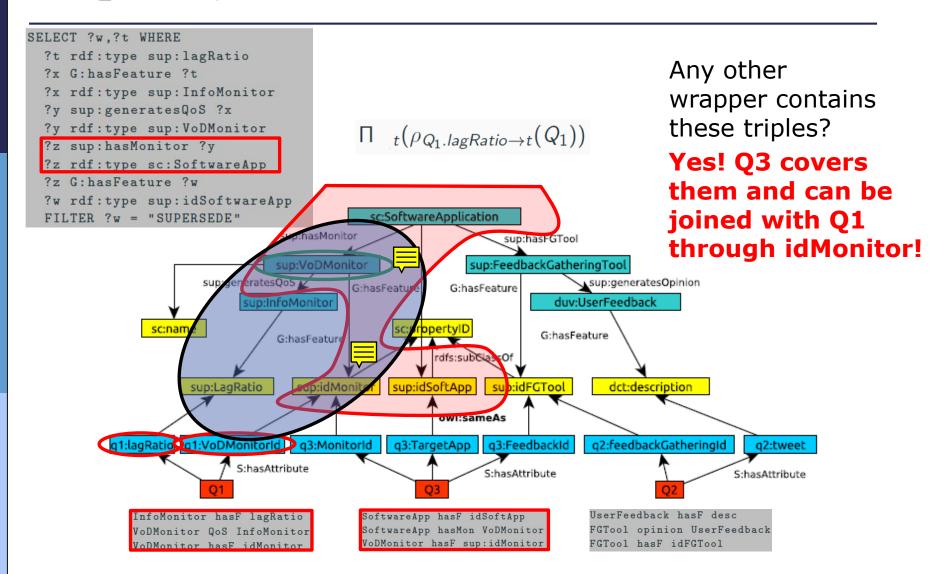
Navigate G from the Feature



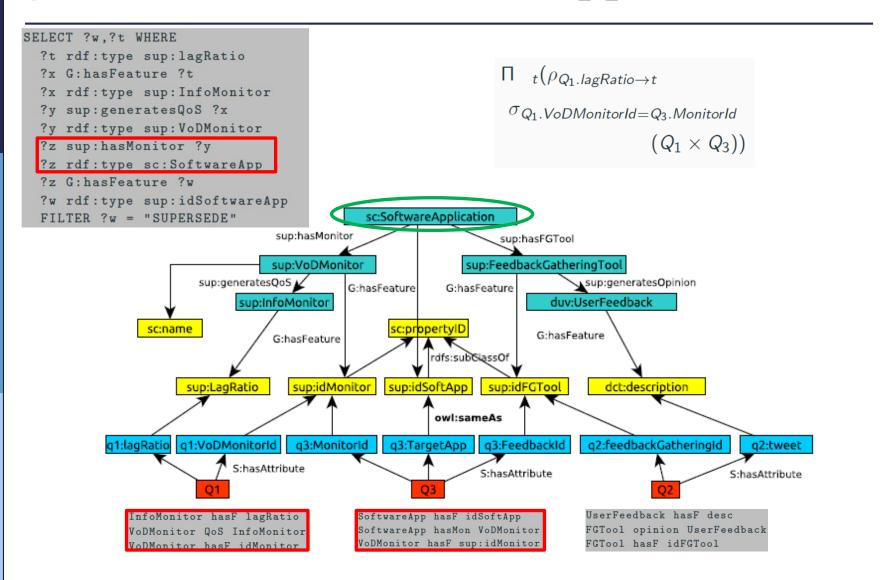
Navigate G from the Feature



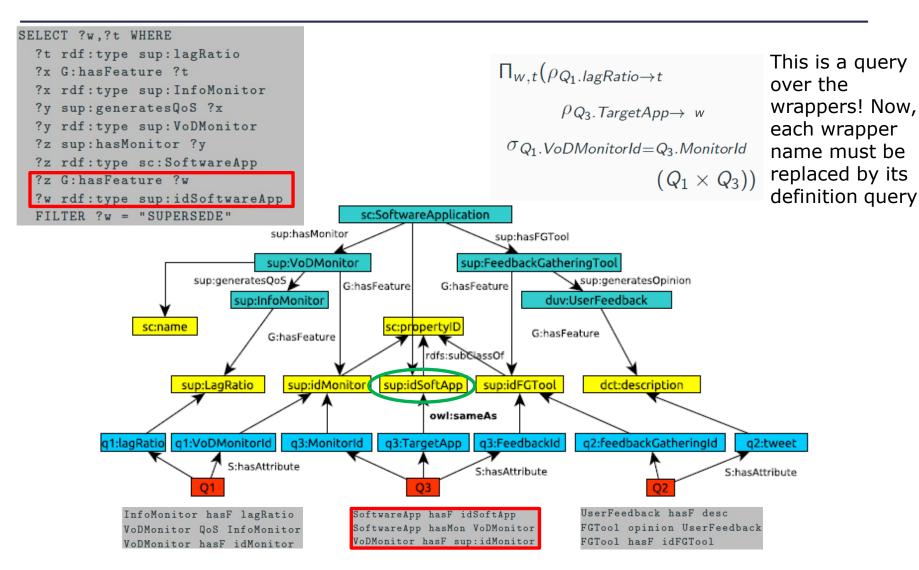
Explore Join Candidates



Join to an Alternative Wrapper



Continue Navigating G

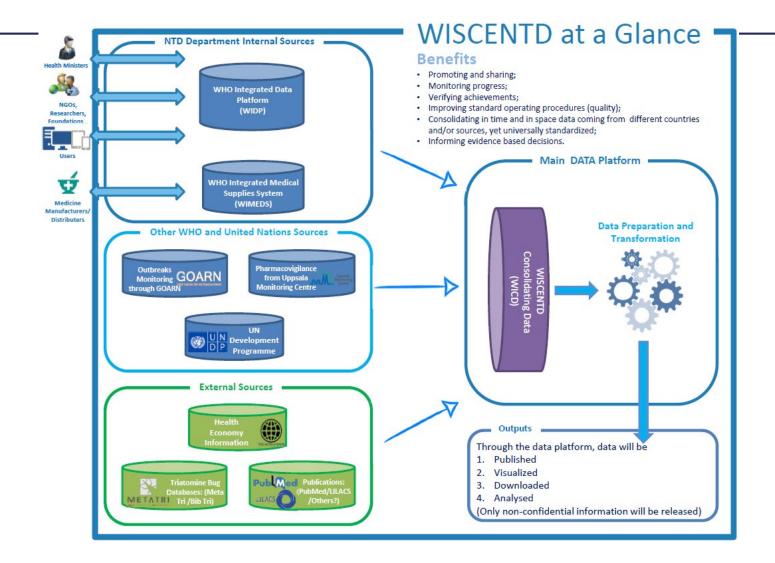


Computational Complexity

- This query rewriting algorithm is:
 - Linear in the size of the subgraph of *G* to navigate
 - Linear in the size of the wrappers mappings
 - Exponential in the number of wrappers that may join
 - Our experiments show that typically Big Data sources have few join points and therefore this exponential complexity is affordable in real cases

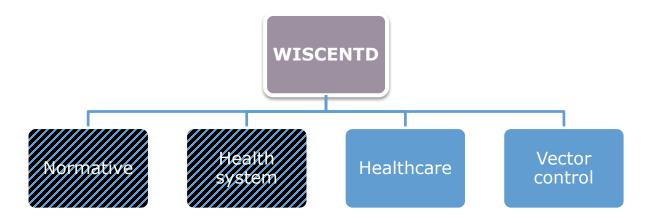


Example of application: The World Health Organisation



Standardisation

- Data has been organized into 4 packages
 - **Healthcare**: to collect patient data
 - Vector control: to collect data on vector control activities
 - Health system: to collect general information on how NTDs are included in the national health systems
 - Normative: to collect information about regulations implemented in to country in order to control and eliminate NTDs

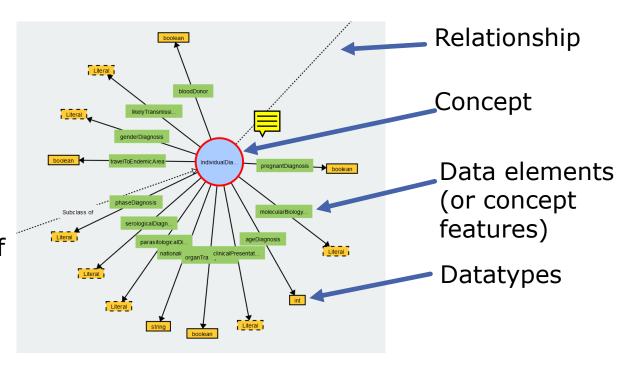


Standardisation

WISCENTD provides a single standardised view of the whole domain

WISCENTD provides a graph-based metaphor representing the domain:

- Concepts
- Data elements of each concept (and their datatypes)
- Relationships between concepts



Master Data: Geographic and Temporal Components

Coordinates



Timestamps



Polygons

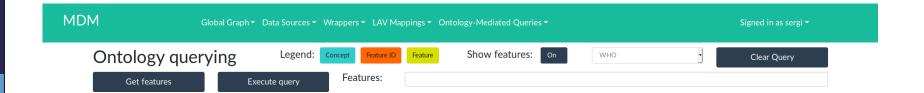


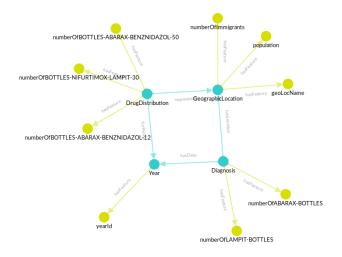
LOCATION

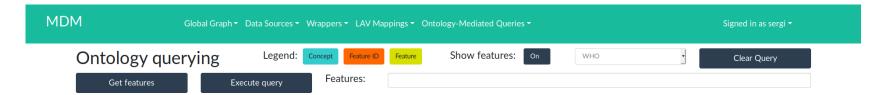
Time periods

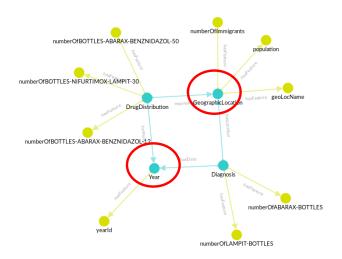
^	August 2013					~	^	September 2013						
Su	Мо	Tu	We	Th	Fr	Sa	Su	Mo	Tu	We	Th	Fr		
				1	2	3	25	26	27	28	29	30	3	
4	5	6	7	8	9	10	1	2	3	4	5	6	7	
11	12	13	14	15	16	17	8	9	10	11	12	13	14	
18	19	20	21	22	23	24	15	16	17	18	19	20	2	
	26	27	28	29	30	31	22	23	24	25	26	27	2	
1	2	3	4	s	6	7	29	30						
	Tuesday, August 13, 2013							Tuesday, August 13, 2013						

TIME



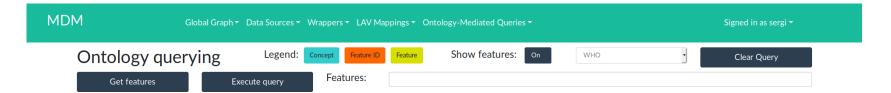


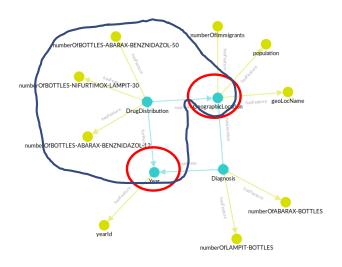




Master data:

geographical and temporal components



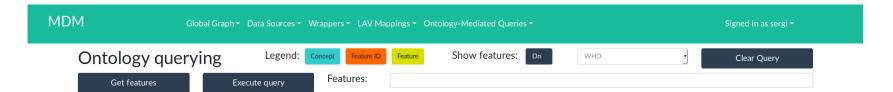


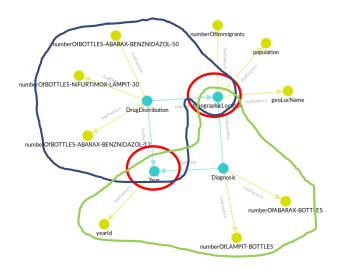
Master data:

geographical and temporal components

WIMEDS:

medicament request and distribution





Master data:

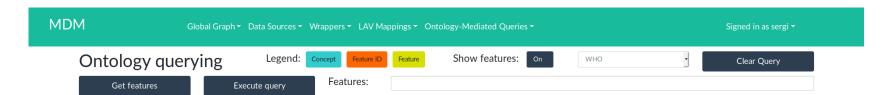
geographical and temporal components

WIMEDS:

medicament request and distribution

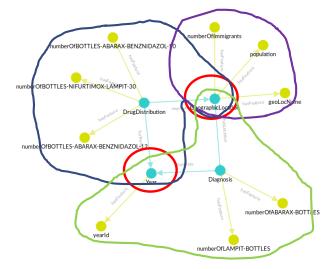
WIDP:

diagnosis and treatment



UN Data:

Health
economics
(two sources:
population and
immigration
data)



Master data:

geographical and temporal components

WIMEDS:

medicament request and distribution

WIDP:

diagnosis and treatment

"I would like to correlate the number of treatments with the population and number of immigrants of a specific greographical area per year"

Find this video in Learn-SQL (video 1)

"I would like to correlate the number of treatments with the number of medicines distributed in a specific greographical area per year.

This information should also include the population and the number of immigrants of that area"

Find this video in Learn-SQL (video 2)

Management: Extending the Ontology

My new data source contains data elements not covered by the attributes of the standardised model. How do I extend it? Find this video in Learn-SQL (video 3)

- My data source contains data that is not covered in the attributes of the standardised model. How do I extend it?
- Great! Now, I want to register a new source providing such data

Find this video in Learn-SQL (video 4)

Management: Querying a New Source

- My data source contains data that is not covered in the attributes of the standardised model. How do I extend it?
- Great! Now, I want to register a new source providing such data
- Good! Now "I would like to see the medicine distribution per medicine sender, per year and geographical area"

Find this video in Learn-SQL (video 5)

A Tool for OMQ

ODIN:

http://www.essi.upc.edu/~snadal/odin.html

- RDFS / OWL (but limited reasoning)
- LAV mappings
- Pay-as-you-go data integration



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Summary

- Graph-based Virtual Data Integration
- Ontology-based Data Access
 - DL-Lite
 - GAV mappings
 - Linking Data to Ontologies
- Ontology-mediated Queries
 - RDFS
 - LAV mappings
 - Sources exposed as wrappers

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