

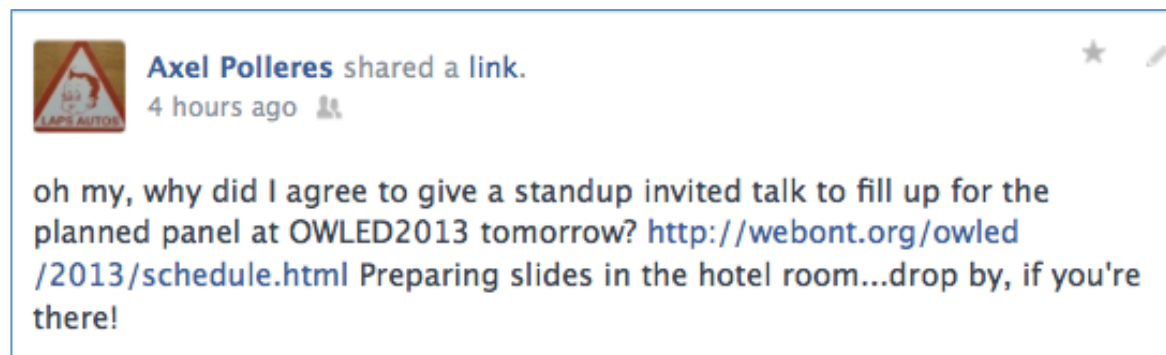


OWL vs. Linked Data: Experiences and Directions

Axel Polleres
Siemens AG Österreich

27/05/2013

Disclaimer: this is not the panel you were looking for ²



Will try to touch upon the following questions:

- Which parts of OWL are used within Linked Data?
- Which parts of OWL2 could be useful for Linked Data ?
- Which reasoning techniques can be applied to handle the scale, messiness, and dynamicity of Linked Data?
- Which reasoning beyond RDFS and OWL is necessary for Linked Data?

I don't have complete solutions to all these questions, but some on how OWL can be applied to Linked Data collected over the past view years.

Special thanks to my co-authors: **Aidan Hogan, Jürgen Umbrich, Stefan Bischof, Andreas Harth, Birte Glimm, Markus Krötzsch**, etc.

Let's start from the beginning...



*“If **HTML** and the **Web** made all the online documents look like one huge **book**,
RDF, **schema** and **inference** languages will make all the data in the world look like
one huge **database**”*
Tim Berners-Lee, *Weaving the Web*, 1999

Great! We have 2013, this should work by now, shouldn't it? Let's try that on google!

Scenario: *“Market research” on the Web, (no FOAF today, sorry)*



“Latest news on NYT about technology companies
with a revenue greater than 10B EUR”



“Which city of Montpellier and Vienna has the higher population density?”



“Latest News on NYT about technology companies with a revenue greater than 10B EUR”

- The data is there!

Wikipedia

New York Times thesaurus and article search API

data.nytimes.com

For the last 150 years, The New York Times has maintained one of the most authoritative news vocabularies ever developed. In 2009, we began to publish this vocabulary as linked open data.

The Data

As of 13 January 2010, The New York Times has published approximately 10,000 subject headings as linked open data under a CC BY license. We provide both RDF documents and a human-friendly HTML versions. The table below gives a breakdown of the various tag types and mapping strategies on data.nytimes.com.

Type	Manually Mapped Tags	Automatically Mapped Tags	Total
People	4,978	0	4,978
Organizations	0	1,592	3,081
Descriptors	498	0	498
			10,467

Browse individual data records: A B C D E F G H I J K L M N O P Q R S T U V W X Y Z

IBM

From Wikipedia, the free encyclopedia

International Business Machines Corporation (NYSE: [IBM](#)), or **IBM**, is an American multinational technology and consulting corporation, with headquarters in Armonk, New York, United States. IBM manufactures and markets computer hardware and software, and offers infrastructure, hosting and consulting services in areas

Siemens

From Wikipedia, the free encyclopedia

For other uses of "Siemens", see *Siemens (disambiguation)*.

Siemens AG (German pronunciation: [ˈzi:məns]) is a German multinational engineering and electronics conglomerate company headquartered in Munich, Germany. It is the largest Europe-based electronics and electrical engineering company.^[2]

Siemens' principal activities are in the fields of industry, energy, transportation and healthcare. It is organized into five main divisions: Industry, Energy, Healthcare, Infrastructure & Cities, and Siemens Financial Services

Siemens AG

Type	Aktiengesellschaft
Traded as	FWB: SIE, NYSE: SI
Industry	Conglomerate
Founded	October 1, 1847 (Berlin)
Revenue	€73.52 billion (2011) ^[1]

SAP AG

SAP AG (ISIN: DE0007164600, FWB: [SAP](#), NYSE: [SAP](#)) is a German multinational software corporation that makes enterprise software to manage business operations and customer relations. Headquartered in Walldorf, Baden-Württemberg, with regional offices around the world, SAP is the market leader in enterprise application software. The company's best-known software products are its enterprise resource planning application ([SAP ERP](#)), its enterprise warehouse solution - SAP Business Warehouse, SAP BusinessObjects software, and most recently, its mobile products and in-memory computing solutions [SAP HANA](#). SAP is one of the largest software companies in the world.

SAP AG

Type	Aktiengesellschaft
Industry	Enterprise software
Revenue	€14.233 billion (2011)

“Which city of Montpellier and Vienna has the higher population density?”

- Again:
The data is there!

Wikipedia

Vienna - Wikipedia, the free encyclopedia - Mozilla Firefox

en.wikipedia.org/wiki/Vienna

Area

- City: 414.85 km²

Population (2011)

- City: 1,714,142 ▲
- Density: 4,002.2/km²

Urban: 1,983,836
Metro: ca. 2,419,000

Eurostat



Urban Audit
(ca. 300 indicators for ~500 European cities)

Score: 0.00, 8.92, 17.83, 17.83

Average: 3.88 High: 17.83 Low: 1.4

Proportion of journeys to work by public transport (rail, metro, bus, tram)

Average: 20.25 High: 41.3 Low: 4.9

You are on page 1 of 4 (40 records)

Rank	City	Score
1	DE1001I	Total resident population
2	EN5101I	Population density: total resident pop. per square km
3	TT1020I	Average length of journey to work by private car (km)
	TT1066I	Length of public transp.network as a prop. of land area
	TT1076I	Length of public transport network per 1000 non
8	Nürnberg (DE)	29.60
9	Hannover (DE)	29.20
10	Köln - Cologne (DE)	26.90

Axel Polleres

or also inside Siemens!

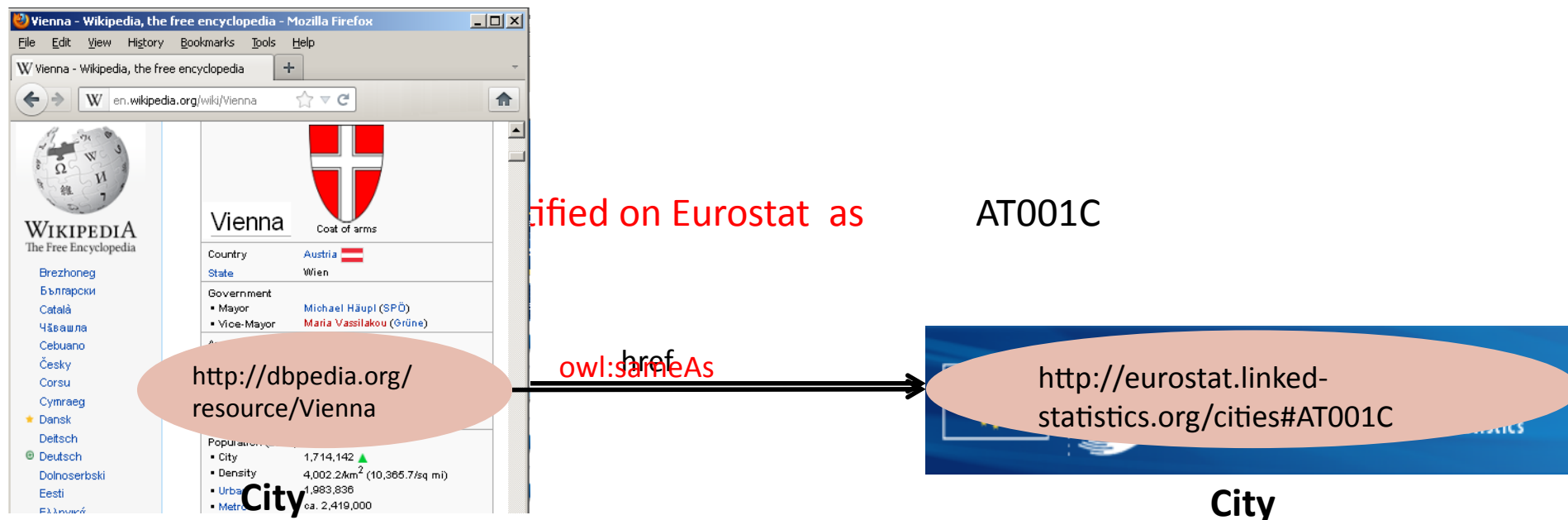
European Green City Index
Assessing the environmental impact of Europe's major cities
A research project conducted by the Economist Intelligence Unit, sponsored by Siemens

City	Score
Oslo	8,71
Copenhagen	8,69
Vienna	7,76
Stockholm	7,61
Amsterdam	7,08
Zurich	6,92
Rome	6,40
Brussels	6,19
Lisbon	5,77
London	5,64

City	Score
=1 Berlin	9,44
=1 Stockholm	9,44
3 Oslo	9,22
4 Copenhagen	9,17
5 Helsinki	9,11
6 Amsterdam	9,01
7 Paris	8,96
8 Vienna	8,62
9 Zurich	8,43
10 London	7,96

City	Score
1 Stockholm	8,81
2 Amsterdam	8,44
3 Copenhagen	8,29
4 Vienna	8,00
5 Oslo	7,92
6 Zurich	7,83
7 Brussels	7,49
8 Bratislava	7,16
9 Helsinki	7,08
=10 Budapest	6,64
=10 Tallinn	6,64

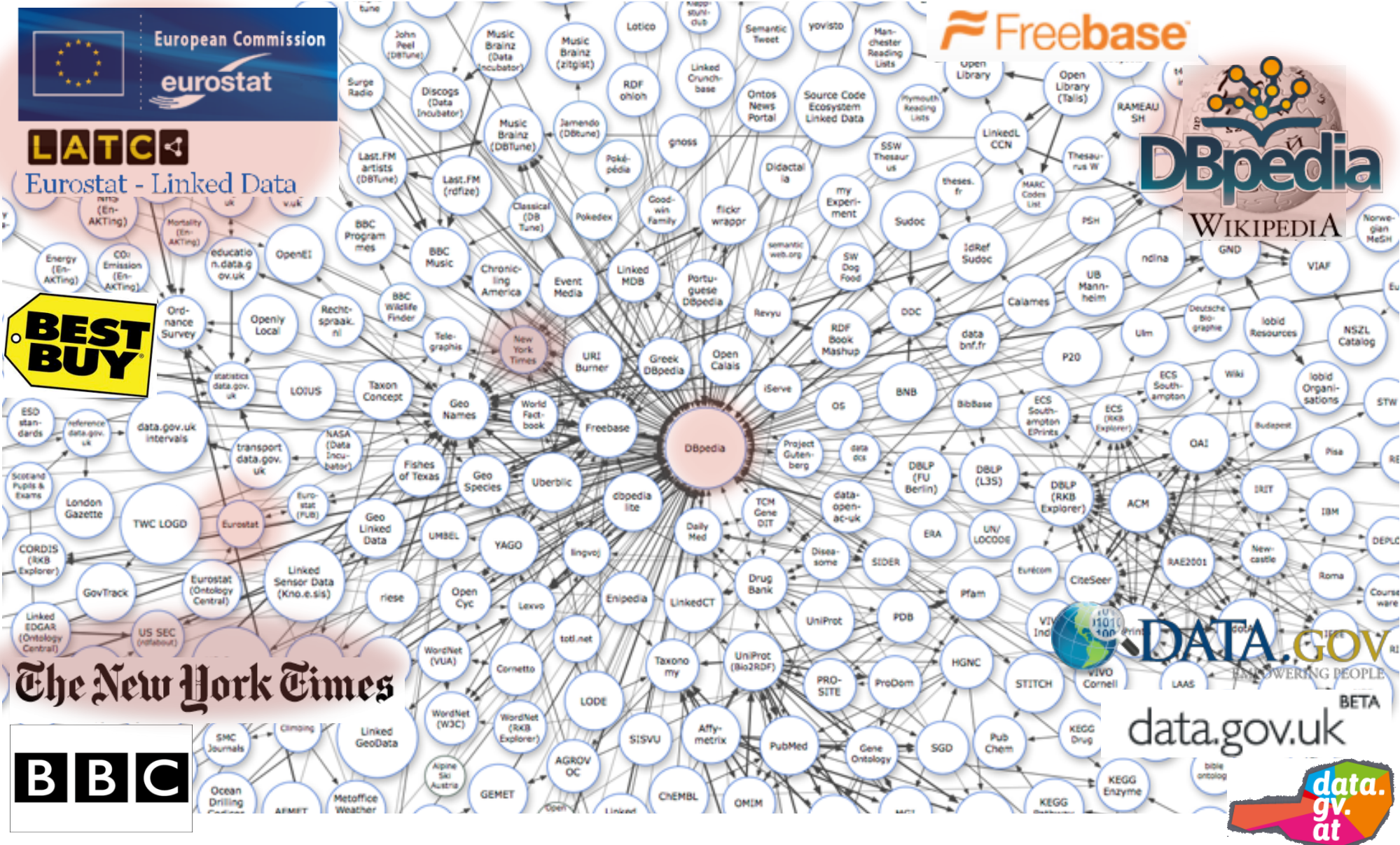
Emerging trend: Linked data



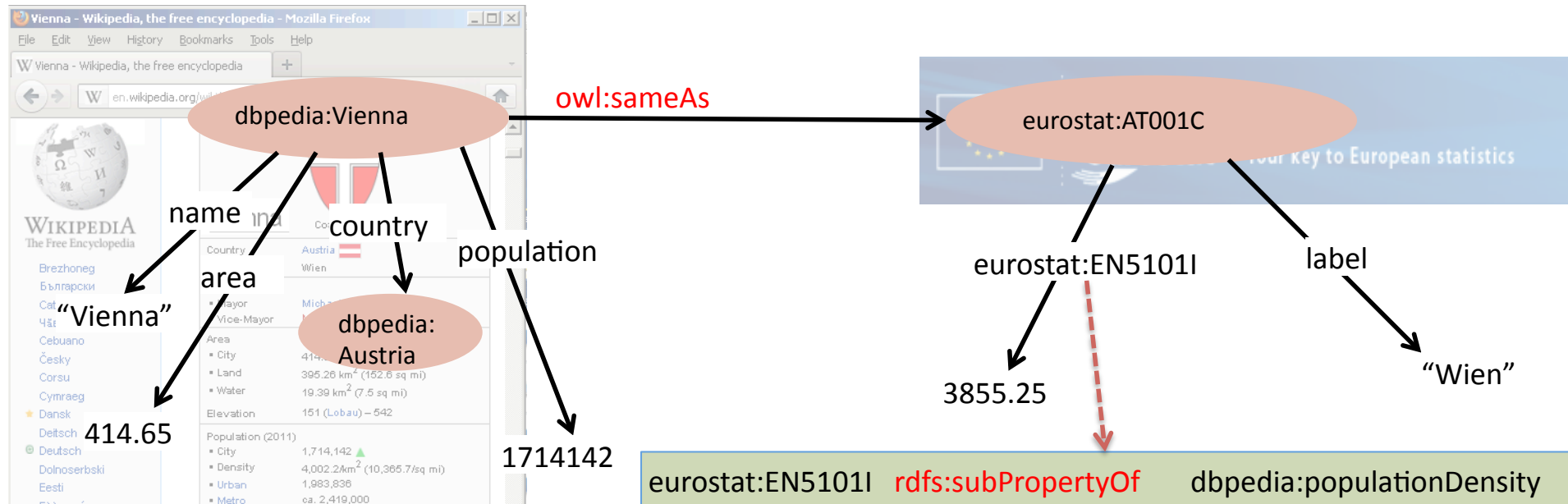
1. Use URIs as names for things
2. Use HTTP URIs so that people can look up those names.
3. When someone looks up a URI, provide useful information, using the standards (RDF, OWL)
4. Include links to other URIs. so that they can discover more things.

<http://www.w3.org/DesignIssues/LinkedData.html>

Linked Data on the Web: Adoption



Example: Using OWL and SPARQL to query Linked data



“Which city of Montpellier and Vienna has the higher population density?”

```
SELECT ?DM ?DV
WHERE {
  dbpedia:Montpellier :populationDensity ?DM .
  dbpedia:Vienna :populationDensity ?DV .
}
```


The Semantic promise...



*“If **HTML** and the Web made all the online documents look like one huge **book**,
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Great! We have 2013, this should work by now, shouldn't it? Let's try that on google!

Scenario: *“Market research” on the Web,*

“Latest news on NYT about technology companies
with a revenue greater than 10B EUR”



```
SELECT * WHERE
{ ?C rdf:type NYT:Org .
  ?C dbpedia:revenue ?R .
  ?C NYT:latestArticle ?A .
  FILTER( ?R > 10000000000 ) }
```



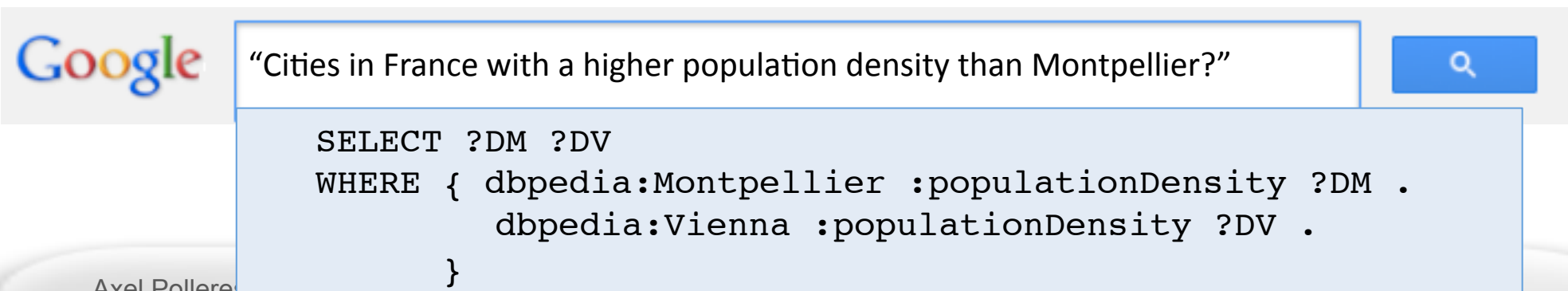
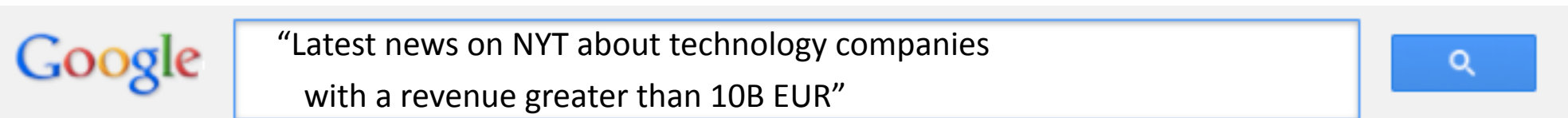
The Semantic promise...



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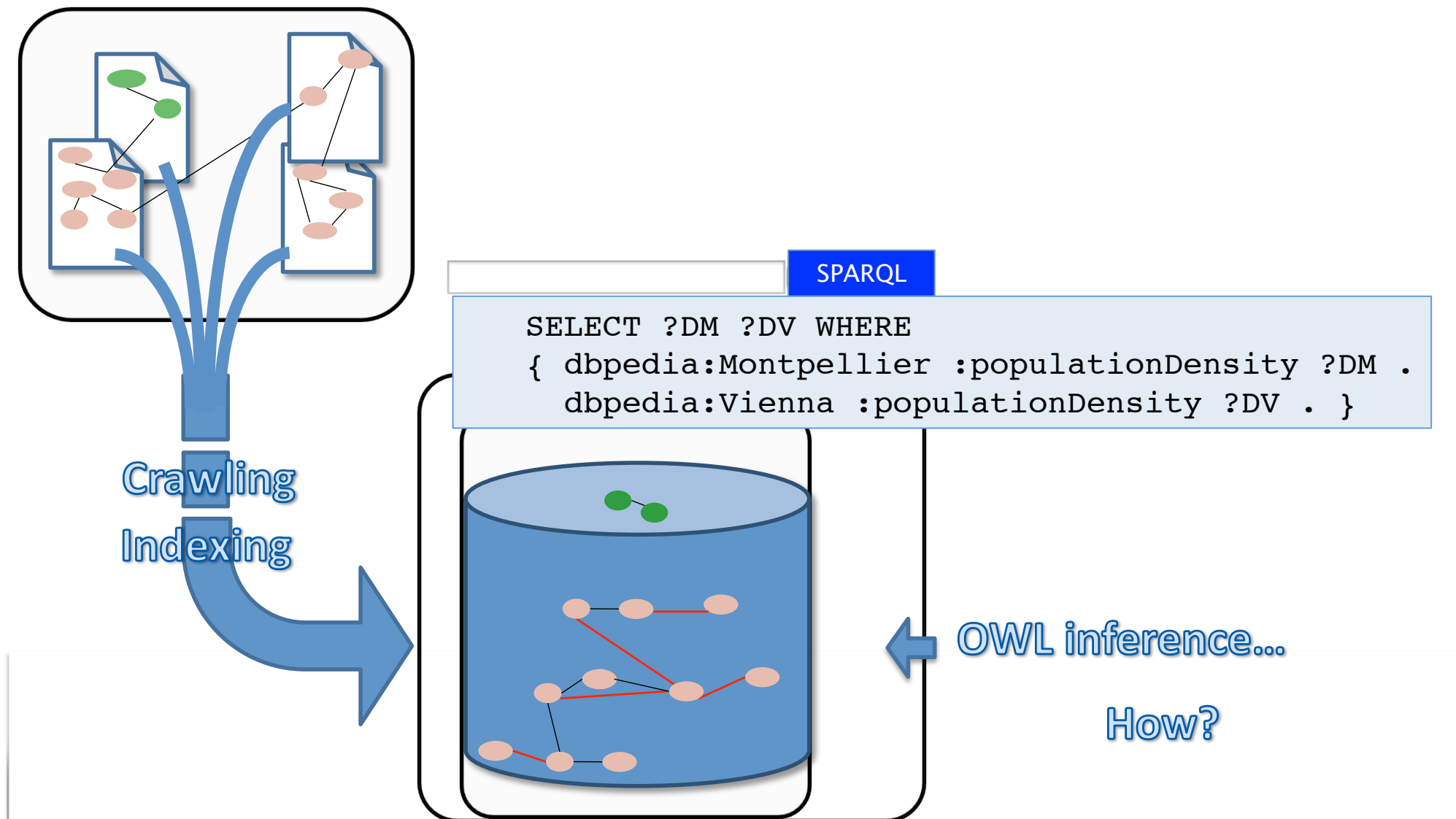
Great! We have 2013, this should work by now, shouldn't it? Let's try that on google!

Scenario: “Market research” on the Web,



So, all we need to do:

- ... is to through together a crawler, an RDF Store, some OWL inference and a SPARQL engine? Unfortunately it's not that easy...



Obstacles/prejudices:

- OWL is too hard to learn?
- OWL is too expensive?

Linked Data publishers...



www.w3.org/TR/2009/REC-owl2-direct-semantics-20091027/

2.2 Interpretations

Given a datatype map D and a vocabulary V over D , an *interpretation* $I = (\Delta_I, \Delta_D, \cdot^C, \cdot^{OP}, \cdot^{DP}, \cdot^I, \cdot^{DT}, \cdot^{LT}, \cdot^{FA})$ for D and V is a 9-tuple with the following structure:

- Δ_I is a nonempty set called the *object domain*.
- Δ_D is a nonempty set disjoint with Δ_I called the *data domain* such that $(DT)^{DT} \subseteq \Delta_D$ for each datatype $DT \in V_{DT}$.
- \cdot^C is the *class interpretation function* that assigns to each class $C \in V_C$ a subset $(C)^C \subseteq \Delta_I$ such that
 - $(owl:Thing)^C = \Delta_I$ and
 - $(owl:Nothing)^C = \emptyset$.
- \cdot^{OP} is the *object property interpretation function* that assigns to each object property $OP \in V_{OP}$ a subset $(OP)^{OP} \subseteq \Delta_I \times \Delta_I$ such that
 - $(owl:topObjectProperty)^{OP} = \Delta_I \times \Delta_I$ and
 - $(owl:bottomObjectProperty)^{OP} = \emptyset$.
- \cdot^{DP} is the *data property interpretation function* that assigns to each data property $DP \in V_{DP}$ a subset $(DP)^{DP} \subseteq \Delta_I \times \Delta_D$ such that
 - $(owl:topDataProperty)^{DP} = \Delta_I \times \Delta_D$ and
 - $(owl:bottomDataProperty)^{DP} = \emptyset$.
- \cdot^I is the *individual interpretation function* that assigns to each individual $a \in V_I$ an element $(a)^I \in \Delta_I$.
- \cdot^{DT} is the *datatype interpretation function* that assigns to each datatype $DT \in V_{DT}$ a subset $(DT)^{DT} \subseteq \Delta_D$ such that
 - \cdot^{DT} is the same as in D for each datatype $DT \in N_{DT}$, and
 - $(rdfs:Literal)^{DT} = \Delta_D$.
- \cdot^{LT} is the *literal interpretation function* that is defined as $(It)^{LT} = (LV, DT)^{LS}$ for each $It \in V_{LT}$, where LV is the lexical form of It and DT is the datatype of It .
- \cdot^{FA} is the *facet interpretation function* that is defined as $(F, It)^{FA} = (F, (It)^{LT})^{FS}$ for each $(F, It) \in V_{FA}$.

The following sections define the extensions of \cdot^{OP} , \cdot^{DT} , and \cdot^C to object property expressions, data ranges, and class expressions.

2.2.1 Object Property Expressions

The object property interpretation function \cdot^{OP} is extended to object property expressions as shown in Table 1.

...OWL IS HARD

(...to learn, to understand, to implement, to compute, to teach, to represent in RDF, to publish, to parse, to use *appropriately*...)

Corollary1: Linked Data publishers only use a little bit of OWL ...

Corollary2: ... they still manage to make mistakes 😊

Corollary1: Linked Data publishers only use a little bit of OWL ...

Proof

How much OWL is used in Linked Data? [LDOW'12]

- Looked at Billion Triple Challenge 2011 Dataset (BTC2011)
 - 2.1 billion quadruples, crawled from...
 - 7.4 million RDF/XML documents, covering...
 - 791 (pay-level) domains
- Count OWL features used in the dataset:
 - Per use
 - Per document
 - Per domain
 - **Can be skewed by data**
- Ranked OWL features using PageRank:
 - Rank documents based on dereferenceable links
 - For each OWL feature, sum the rank of documents using it
 - **Intuition: Approximates probability of encountering an OWL feature during a random walk of the data**

Results of ranking (see LDOW'12 paper details)

1	<code>rdf:Property</code>	5.74E-1
2	<code>rdfs:range</code>	4.67E-1
3	<code>rdfs:domain</code>	4.62E-1
4	<code>rdfs:subClassOf</code>	4.60E-1
5	<code>rdfs:Class</code>	4.45E-1
6	<code>rdfs:subPropertyOf</code>	2.35E-1
7	<code>owl:Class</code>	1.74E-1
8	<code>owl:ObjectProperty</code>	1.68E-1
9	<code>rdfs:Datatype</code>	1.68E-1
10	<code>owl:DatatypeProperty</code>	1.65E-1
11	<code>owl:AnnotationProperty</code>	1.60E-1
12	<code>owl:FunctionalProperty</code>	9.18E-2
13	<code>owl:equivalentProperty</code>	8.54E-2
14	<code>owl:inverseOf</code>	7.91E-2
15	<code>owl:disjointWith</code>	7.65E-2

Results of ranking (see LDOW'12 paper details)

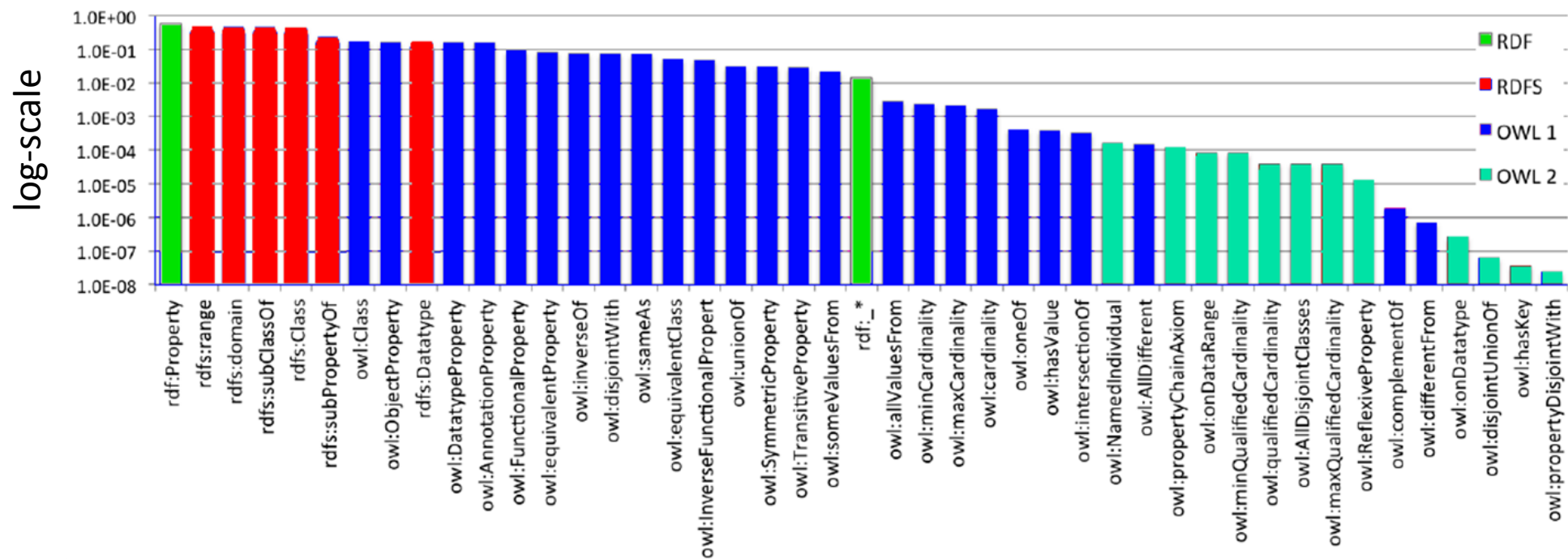
...

16	<code>owl:sameAs</code>	7.29E-2
17	<code>owl:equivalentClass</code>	5.24E-2
18	<code>owl:InverseFunctionalProperty</code>	4.79E-2
19	<code>owl:unionOf</code>	3.15E-2
20	<code>owl:SymmetricProperty</code>	3.13E-2
21	<code>owl:TransitiveProperty</code>	2.98E-2
22	<code>owl:someValuesFrom</code>	2.13E-2
23	<code>rdf: *</code>	1.42E-2
24	<code>owl:allValuesFrom</code>	2.98E-3
25	<code>owl:minCardinality</code>	2.43E-3
26	<code>owl:maxCardinality</code>	2.14E-3
27	<code>owl:cardinality</code>	1.75E-3
28	<code>owl:oneOf</code>	4.13E-4
29	<code>owl:hasValue</code>	3.91E-4
30	<code>owl:intersectionOf</code>	3.37E-4
31	<code>owl:NamedIndividual</code>	3.37E-4

How much OWL is used in Linked Data? [LDOW'12]

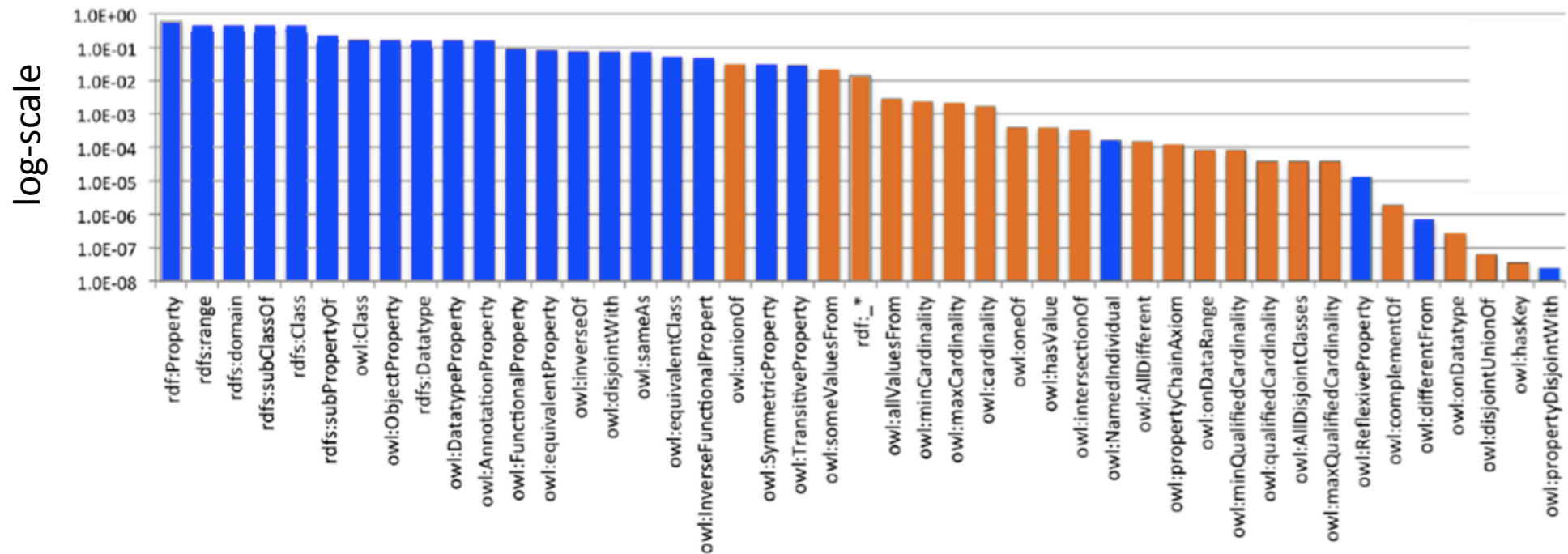
- RDF Schema features amongst the most prominently used
- OWL 2 features not yet used prominently

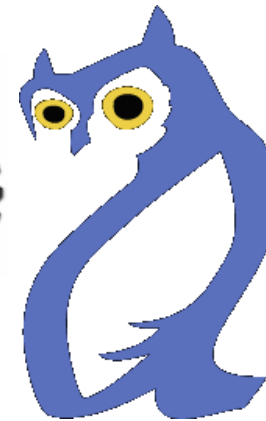
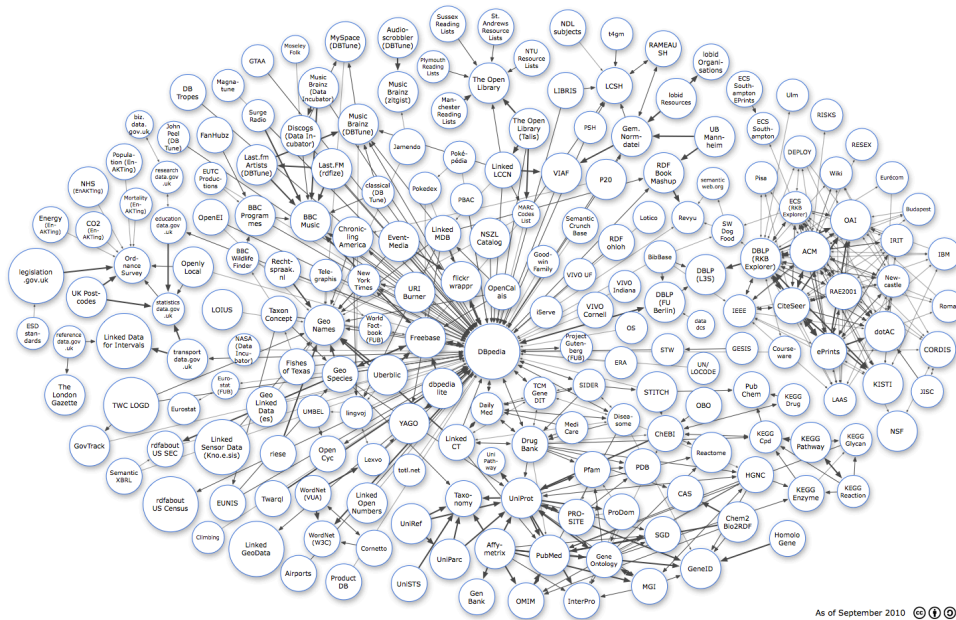
RDF | RDFS | OWL | OWL 2



Main insight:

- RDF Schema features amongst the most prominently used
- OWL 2 features not yet used prominently
- Most *used* features can be implemented efficiently using **parallelizable, rule-based inference.** [IJSWIS'09]





Intuition: **MAYBE LINKED DATA ONLY NEEDS
A LITTLE OWL...**

(...for now)

[Hendler, '98]

How much OWL is used in Linked Data? [LDOW'12]

- RDF Schema features amongst the most prominently used
- OWL 2 features not yet used prominently

In all fairness:

This is only a snapshot...

... some OWL2 features could be quite useful to enrich current Linked Data vocabularies, see also [RR'09]

We presented this 3 years ago at ESWC... still only little OWL2 adoption in LOD vocabs since then...

OWL2



ESWC2010

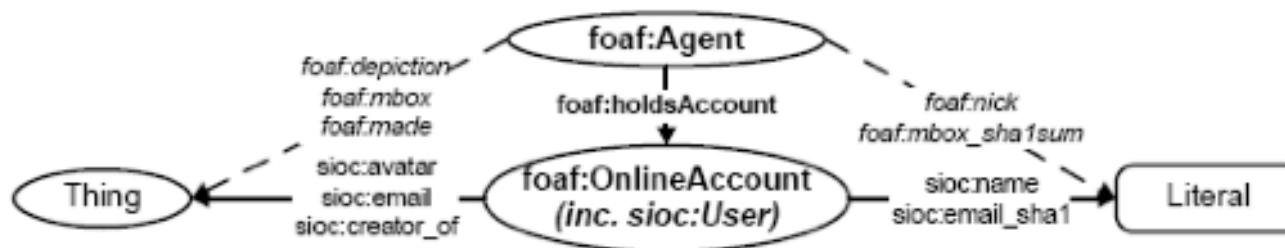
Common ontologies on the Web don't use it a lot as of yet...

... but adds interesting functionality, potentially useful for Web ontologies, e.g.

- PropertyChains

- E.g. could be useful to tie `sioc:name` and `foaf:nick` via `foaf:holdsAccount`:

```
foaf:nick owl:propertyChainAxiom (foaf:holdsAccount sioc:name)
```



Corollary2: ... publishers still manage to make mistakes.

Examples...

Inconsistencies/wrong inferences on Web Data

4 main conjectures for possible reasons:

- Publishers deliberately publish spoilt data (“SPAM”)
- Opinions differ
- “URI-sense” ambiguities
- **Accidentally** wrong/inconsistent

Least common?



Most common?

Publishers deliberately publish spoilt data (“SPAM”)

- Examples:
 - a owl:differentFrom a .
 - foaf:knows rdfs:subPropertyOf owl:sameAs .
 - <http://www.polleres.net/nasty.rdf>
- Can occur for “testdata” being published, deliberate SPAM might become an issue, as the SW grows!

Opinions differ

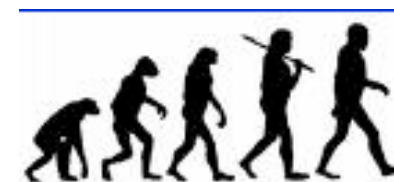
- Fictitious Example Ontology:

Originofthings.example.org:

```
o1:surpremePower owl:disjointWith o1:naturalPhenomenom.  
o1:originsFrom rdf:type owl:functionalProperty.  
o1:god rdf:type o1:surpremePower.  
o1:evolution rdf:type o1:naturalPhenomenom.
```

darwin.example.org:

```
ex:mankind o1:originsFrom o1:evolution .
```



creationism.example.org:

```
ex:mankind o1:originsFrom o1:god
```



FlyingSpaghettimonster.org

```
fsm:theSpaghettiMonster rdf:type surpremePower.  
ex:mankind o1:originsFrom fsm:theSpaghettiMonster.
```



“URI-sense” ambiguities

```
<http://www.polleres.net>  
  foaf:knows <http://apassant.net>
```

i.e., why do I have to use a different URI for myself and my homepage?

Many people don't understand/like this and make mistakes.

But is this really a mistake or a design error?

Accidentally inconsistent data

This appeared a while ago in our crawls:

Source1 (faulty):

```
TimBL foaf:homepage <http://www.w3.org>
```

```
TimBL rdf:type foaf:Person.
```

W3.org:

```
W3C foaf:homepage <http://www.w3.org>
```

```
W3C rdf:type foaf:Organisation.
```

Did occur in our Web crawls at some point, sometimes people don't have the right semantics in mind!

- Suspiciously resembles problems with e.g. flawed HTML ... browsers, normal search engines still have to deal with it...

Accidentally wrong (non-inconsistent data)

- FOAF Ontology:

```
foaf:mbox rdf:type owl:InverseFunctionalProperty .
```

- Careless FOAF exporters produce something like this for any empty email address:

```
ex:alice foaf:mbox "mailto:"
```

```
ex:bob foaf:mbox "mailto:"
```

...

Consequence: InverseFunctionalProperty reasoning on Web Data potentially equates too many things! Dangerous!

Not all of these were real examples, of course, for a more systematic analysis of Linked Data conformance, see [JWS'12a,LDOW'10]

BTW, Didn't even mention ill-typed datatype literals here (very common...)

RDF-ALERTS

Your mission, should you decide to accept it, would be to make the Semantic Web clean ...

<http://swse.deri.org/RDFAlerts/>

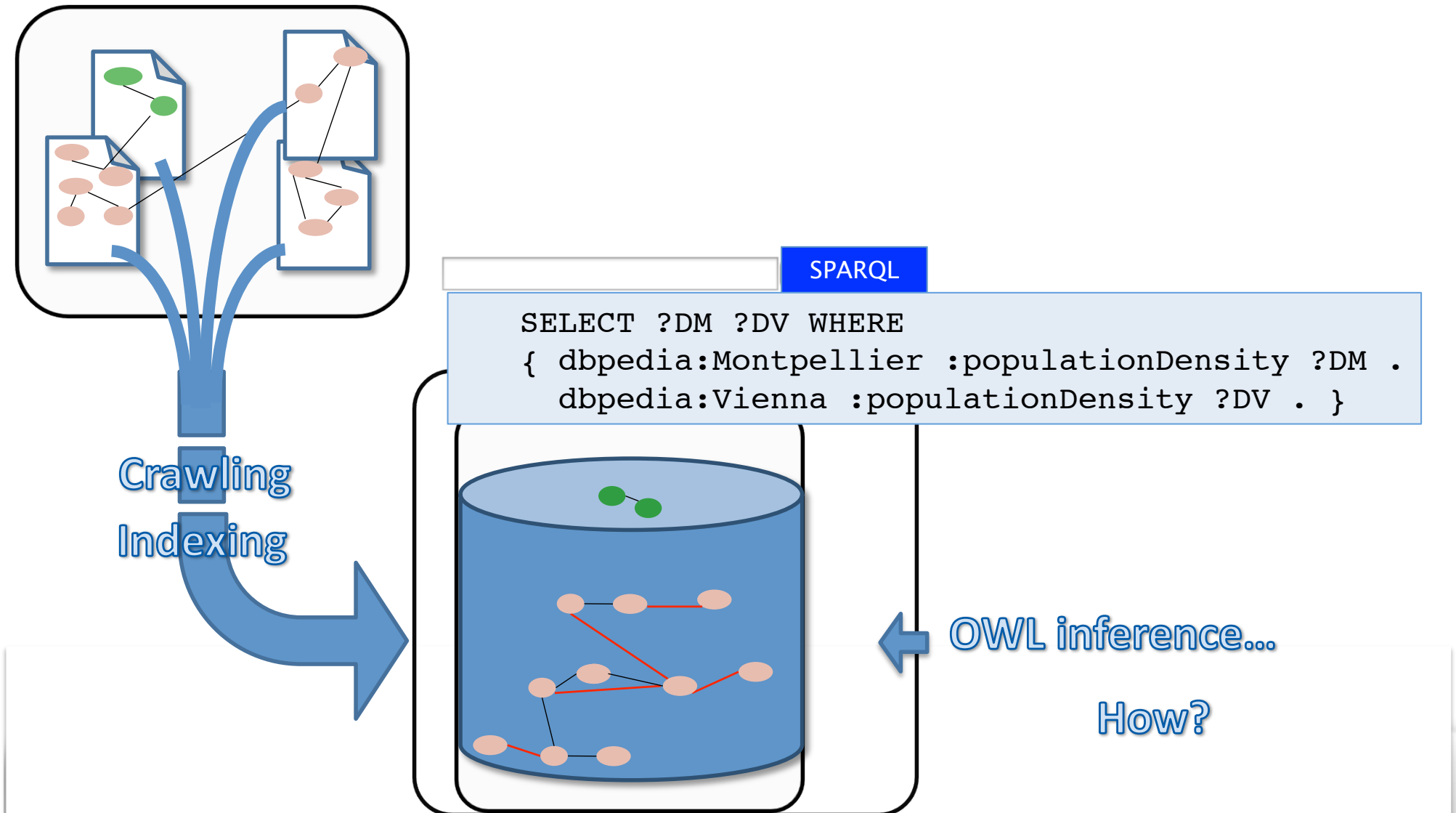
RDF-Alerts

URL:

Please note that we currently only validate RDF/XML, and that validation may take 1-2 minutes due to live crawling of data.

note	error retrieving http://skype.com/ - http://skype.com/ did not return Content-Type application/rdf+xml
okay	retrieved data
warning	could not find a definition for Class http://skype.com/... term does not dereference to an RDF vocabulary description?
warning	could not find a definition for Property http://xmlns.com/foaf/0.1/acountName... term does not dereference to an RDF vocabulary descr
warning	could not find a definition for Property http://xmlns.com/foaf/0.1/acountName... term does not dereference to an RDF vocabulary descr
warning	could not find a definition for Property http://xmlns.com/foaf/0.1/acountName... term does not dereference to an RDF vocabulary descr
okay	finished validation

Conclusion: This won't scale to the Web...



How to possibly avoid nasty inferences?

dbpedia:Vienna

- name: "Vienna"
- area: 414.65
- country: dbpedia:Austria
- population: 1714142

owl:sameAs



eurostat:AT001C

- eurostat:EN5101I: 3855.25
- label: "Wien"

eurostat:EN5101I rdfs:subPropertyOf dbpedia:populationDensity

fakedbpedia.org

dbpedia:populationDensity

owl:sameAs

rdfs:subPropertyOf

Scalable Authoritative OWL Reasoning [IJSWIS'09] to the rescue!

... Goal: make inference **robust & scalable** to Web data.

```
SELECT ?DM ?DV
WHERE {
  dbpedia:Montpellier :populationDensity ?DM .
  dbpedia:Vienna :populationDensity ?DV .
}
```

Scalabe Reasoning: Focus on Rules without ABox joins

DL Syntax	Rule
G0 : NO A-BOX PATTERNS IN ANTECEDENT	
$\{o_i \dots o_n\}$	$\text{?C} : \text{oneOf} (?o_1 \dots ?o_n) . \Rightarrow ?o_1 \dots ?o_n \text{ a ?C} .$
G1 : ONE A-BOX PATTERN IN ANTECEDENT	
$C \sqsubseteq D$	$\text{?C} \text{ rdfs:subClassOf ?D} . ?s \text{ a ?C} . \Rightarrow ?s \text{ a ?D} .$
$C \equiv D$	$\text{?C} \text{ equivalentClass ?D} . ?s \text{ a ?C} . \Rightarrow ?s \text{ a ?D} .$ $\text{?C} \text{ equivalentClass ?D} . ?s \text{ a ?D} . \Rightarrow ?s \text{ a ?C} .$
$P \sqsubseteq Q$	$\text{?P} \text{ rdfs:subPropertyOf ?Q} . ?s \text{ ?P} ?o . \Rightarrow ?s \text{ ?Q} ?o .$
$P \equiv Q$	$\text{?P} \text{ equivalentProperty ?Q} . ?s \text{ ?P} ?o . \Rightarrow ?s \text{ ?Q} ?o .$ $\text{?P} \text{ equivalentProperty ?Q} . ?s \text{ ?Q} ?o . \Rightarrow ?s \text{ ?P} ?o .$
$P \equiv P_0^-$	$\text{?P} \text{ inverseOf ?Q} . ?s \text{ ?P} ?o . \Rightarrow ?o \text{ ?Q} ?s .$ $\text{?P} \text{ inverseOf ?Q} . ?s \text{ ?Q} ?o . \Rightarrow ?o \text{ ?P} ?s .$
$\top \sqsubseteq \forall P^- . C$	$\text{?P} \text{ rdfs:domain ?C} . ?s \text{ ?P} ?o . \Rightarrow ?s \text{ a ?C} .$
$\top \sqsubseteq \forall P . C$	$\text{?P} \text{ rdfs:range ?C} . ?s \text{ ?P} ?o . \Rightarrow ?o \text{ a ?C} .$
$P \equiv P^-$	$\text{?P} \text{ a :SymmetricProperty} . ?s \text{ ?P} ?o . \Rightarrow ?o \text{ ?P} ?s .$
$\exists P . x$	$\text{?C} \text{ :hasValue ?x; :onProperty ?P} . ?y \text{ ?P} ?x . \Rightarrow ?y \text{ a ?C} .$ $\text{?C} \text{ :hasValue ?x; :onProperty ?P} . ?y \text{ a ?C} . \Rightarrow ?y \text{ ?P} ?x .$
$C_1 \sqcup \dots \sqcup C_n$	$\text{?C} \text{ :unionOf} (?C_1 \dots ?C_i \dots ?C_n) . ?x \text{ a ?C}_i . \Rightarrow ?x \text{ a ?C} .$
$(\geq 1P)$	$\text{?C} \text{ :minCardinality } 1; \text{ :onProperty ?P} . ?x \text{ ?P} ?y . \Rightarrow ?x \text{ a ?C} .$
$C_1 \sqcap \dots \sqcap C_n$	$\text{?C} \text{ :intersectionOf} (?C_1 \dots ?C_n) . ?y \text{ a ?C} . \Rightarrow ?y \text{ a ?C}_1, \dots, ?C_n .$
$C_1 \sqcap \dots \sqcap C_n$	$\text{?C} \text{ :intersectionOf} (?C_1) . ?y \text{ a ?C}_1 . \Rightarrow ?y \text{ a ?C} .$

From [IJSWIS'09] The rules applied including statements considered to be T-Box, elements which must be **authoritatively** spoken for (including for **bnode OWL abstract syntax**), and output count.

Abox inferences only! Can be distributed! Several Optimizations in [ISWC2010]

Authoritative Reasoning

- Document **D** authoritative for concept **C** iff:
 - **C** not identified by a URI
 - OR
 - De-referenced URI of **C** coincides with or redirects to **D**
 - FOAF spec authoritative for `foaf:Person` ✓
 - MY Ont not authoritative for `foaf:Person` ✗
- Only allow extension in authoritative documents
 - `my:Person rdfs:subClassOf foaf:Person . (MY Ont) ✓`
- **BUT:** Reduce obscure memberships
 - `foaf:Person rdfs:subClassOf my:Person . (MY Ont) ✗`
- Similarly for other T-Box statements.
- In-memory T-Box stores authoritative values for rule execution



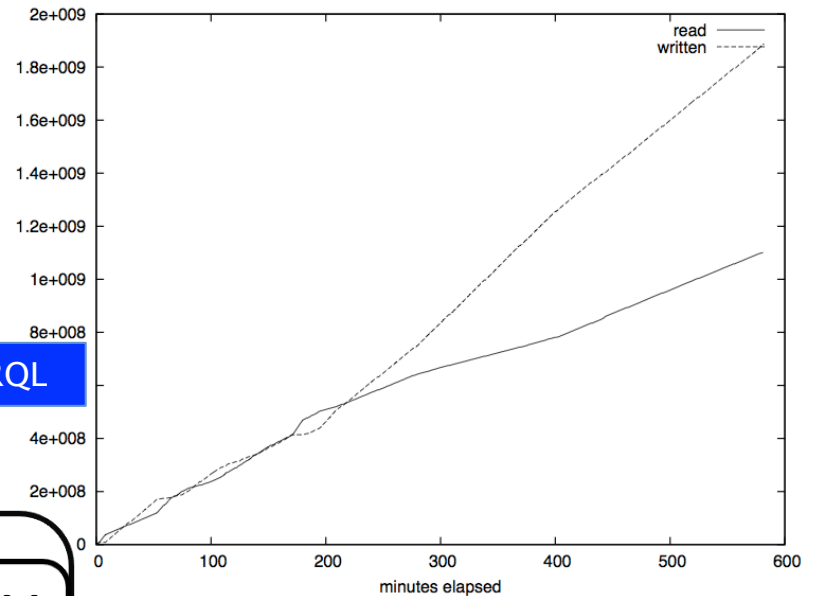
Ontology "Hijacking"



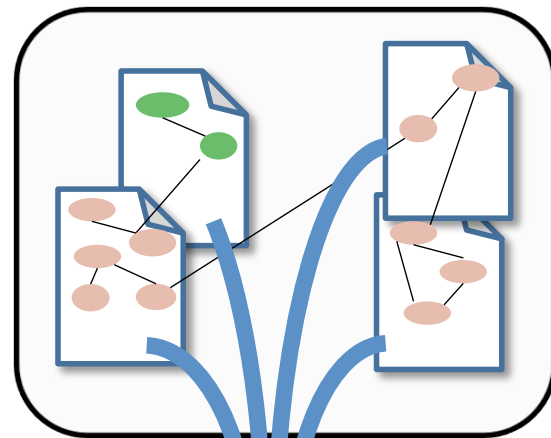
A Semantic Web Search Engine

- ... based on these insights we have implemented SWSE [JWS'11]

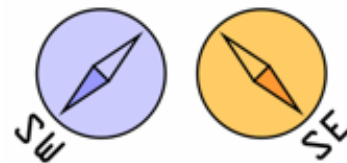
SAOR makes inference **robust & scalable** to Web data:



(on a single machine, goes down near linear in parallel mode)

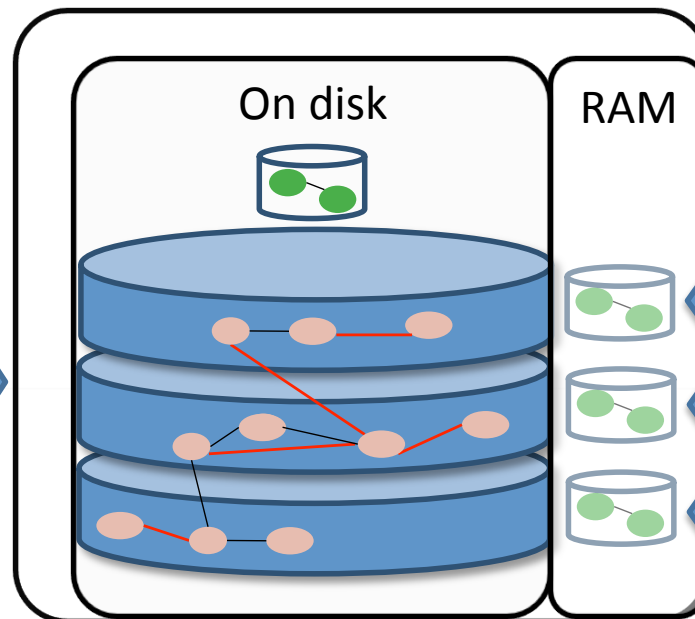


Crawling
Indexing



Semantic Web Search Engine

SPARQL



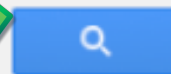
Scalable
Authoritative
OWL Reasoner
(SAOR)

Are we done?

Similar in spirit to SAOR: “Quarantined Reasoning”, cf. [RR’11]



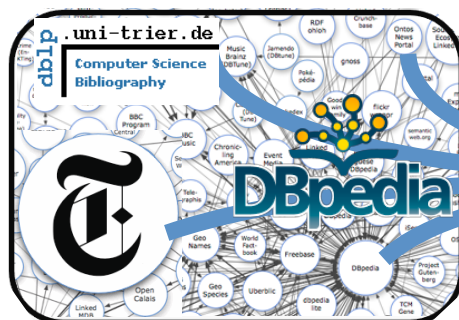
“Which city of Montpellier and Vienna has the higher population density?”



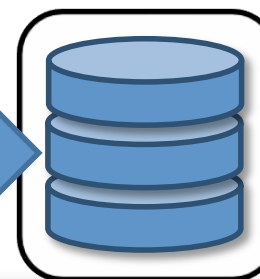
“Latest news on NYT about technology companies with a revenue greater than 10B EUR”



```
SELECT * WHERE
{ ?C rdf:type NYT:Org .
  ?C dbpedia:revenue ?R .
  ?C NYT:latestArticle ?A .
  FILTER( ?R > 10000000000 ) }
```



Crawling
Indexing
Reasoning



Can't keep up with changes!

Idea: “Live”-Linked Data Querying

- Basic idea (originally proposed by Hartig & Bizer)

- Start with URIs in a SPARQL query
- Interleave crawl + query processing

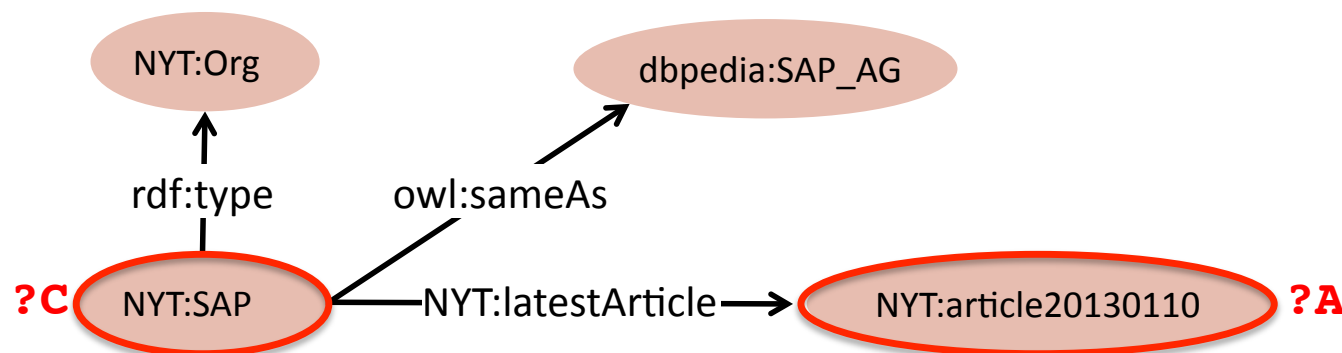
```

SELECT * WHERE
{ ?C rdf:type NYT:Org .
  ?C dbpedia:revenue ?R .
  ?C NYT:latestArticle ?A.
  FILTER (?R > €10000000000 ) }
  
```

HTTP GET NYT:Org

→ HTTP GET NYT:SAP

→ HTTP GET NYT:article20130128_1



**Stop. No new
query
relevant
results found**

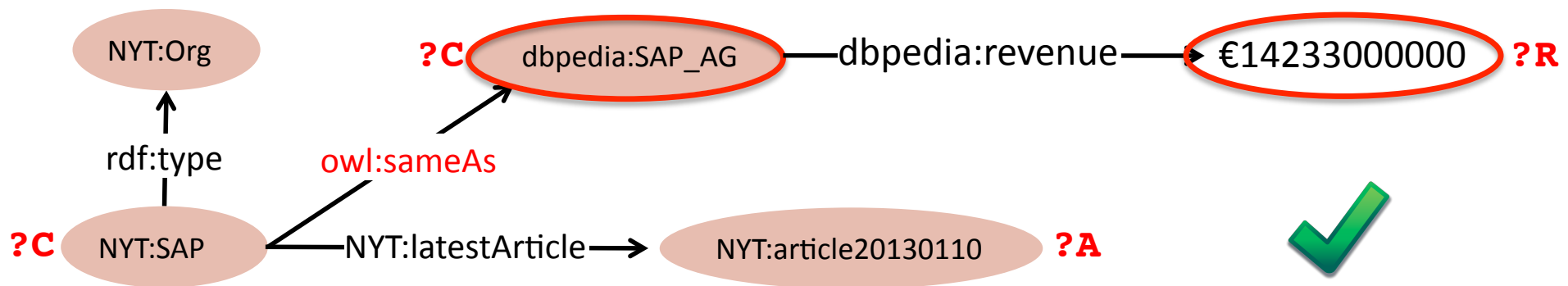
Improved “Live”-Linked Date Querying

- Our approach [RR’12]
- Basic idea:
 - Start with URIs in a SPARQL query
 - **Interleave crawl + query processing**
 - **Lightweight OWL Reasoning**

```

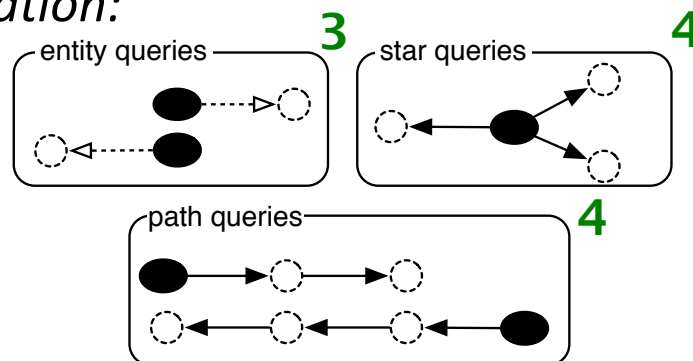
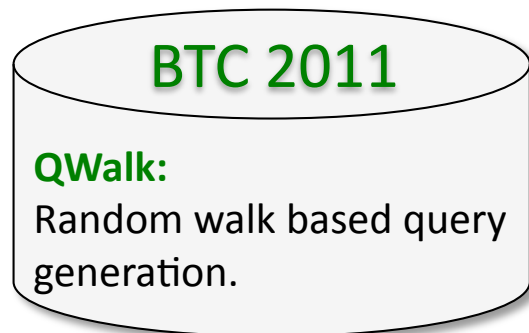
SELECT * WHERE
{ ?C rdf:type NYT:Org .
  ?C dbpedia:revenue ?R .
  ?C NYT:latestArticle ?A.
  FILTER (?R > €10000000000) }
    
```

HTTP GET NYT:Org → HTTP GET NYT:SAP → HTTP GET dbpedia:SAP



Again: Does this work “in the wild”?

1) Benchmark generation:



1100 queries
100 each for 11 “typical”
shapes

2) Results:

- **RDFS** reasoning: *many queries with average impact*
 - In 8/11 query classes more than 50% avg. result increase
 - in 4/11 query classes a time increase of 50%
- **OWL** sameAs: *view queries with high impact*
 - In 2/11 query classes more than 50% avg. result increase
 - In 2/11 query classes a time increase of 50%

Overall our reasoning extensions improved in 8 of 11 query classes the average result/time ratio (throughput)

Are we done?

“Which city of Montpellier and Vienna has the higher population density?”



“Latest news on NYT about technology companies with a revenue greater than 10B EUR”

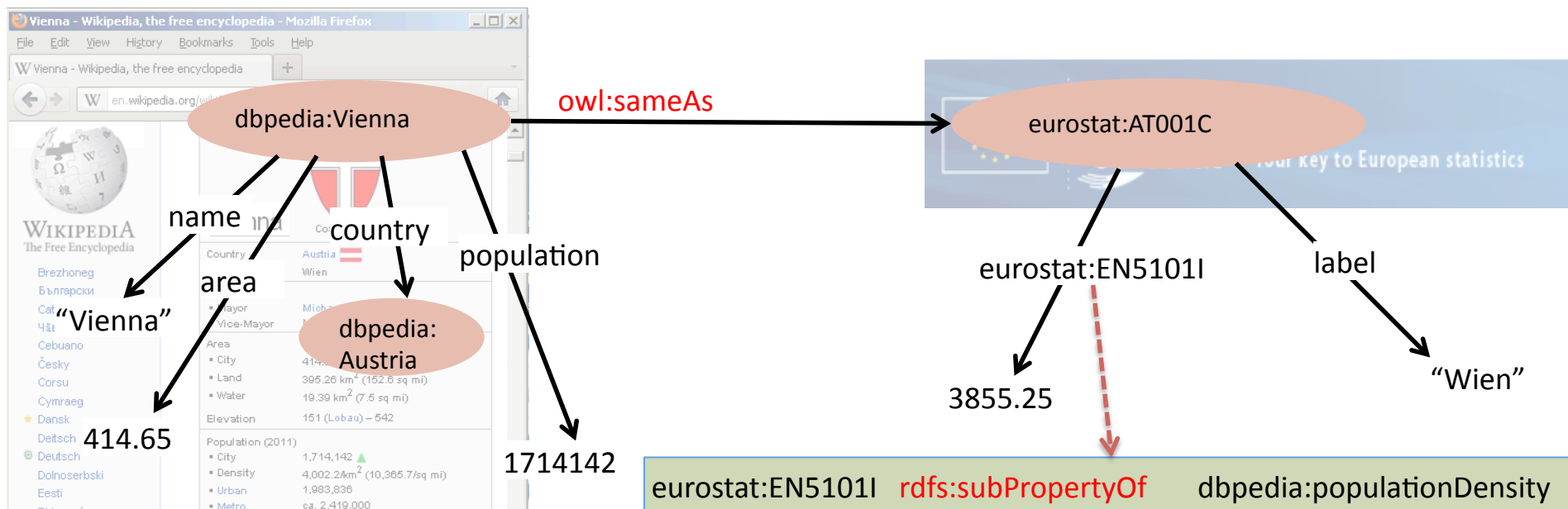


Depends on the point of view...

... e.g.: What is the sweetspot between **centralised & decentralised/live** Linked Data querying approaches?

Plus...

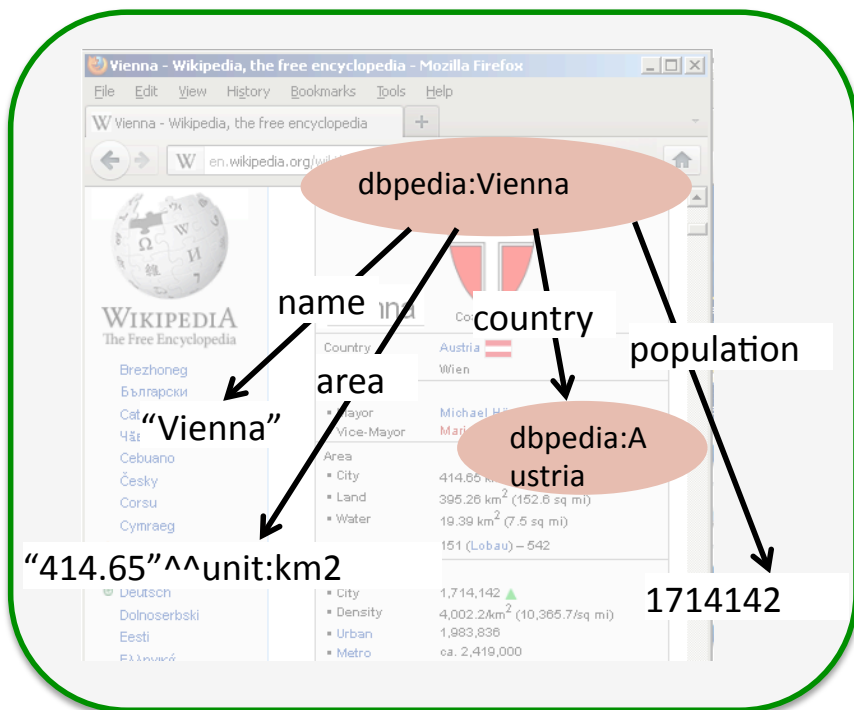
... Actually, we were cheating:



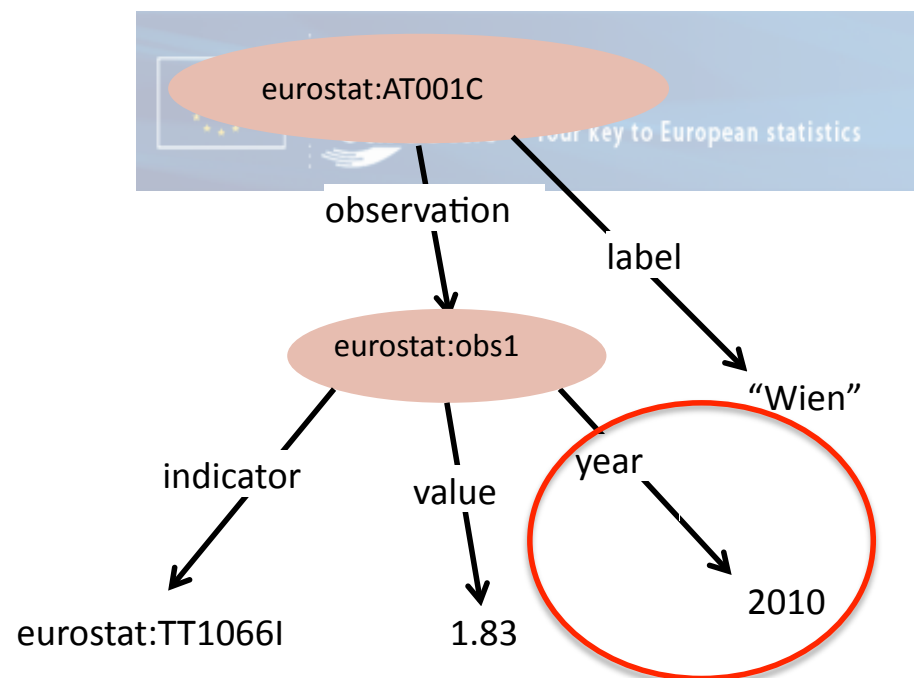
“Which city of Montpellier and Vienna has the higher population density?”

```
SELECT ?DM ?DV
WHERE {
  dbpedia:Montpellier :populationDensity ?DM .
  dbpedia:Vienna :populationDensity ?DV .
}
```

Example: Actually, we were cheating...



Data looks rather like this...



“Which city of Montpellier and Vienna has the higher population density?”

Can't we still answer this query? Just using the Dbpedia data?

We know: **dbpedia:opulationDensity = dbpedia:population / dbpedia:area**

Another example is unit conversion from this morning:

$$\text{elephantWeight} = \text{elphantWeightImperial} * 2.204$$

or

$$\text{priceUSD} = \text{priceEUR} * 1.2850$$

We call these “**propery equation**” ... stay tuned for the main track [ESWC'13]

Take-home messages:

Lots of interesting real-world Datasets out there as Linked Data!

- can benefit from OWL and Reasoning
- (even more Open Data in other formats, BTW)

Structured Query answering over Linked Data needs:

- **Robust & Scalable Reasoning** (à la SAOR)
- Both **centralised & decentralised** approaches
- Maybe more than OWL (e.g. something like property equations)?

Vienna

Population density 2012

4002.2 p/km² (Source: <http://dbpedia.org/>)

Population density 2004

3855.25 p/km²
(Source: <http://www.urbandaudit.org/>)

Population density 2001

3735.24 p/km²
(Source: <http://www.urbandaudit.org/>)

Population density 1999

3735.2 p/km² (Source: <http://eurostat.linked-statistics.org/>)



References:

- [RR'12] Jürgen Umbrich, Aidan Hogan, Axel Polleres, Stefan Decker: Improving the Recall of Live Linked Data Querying through Reasoning. *In 6th International Conference on WebReasoning and Rule Systems (RR2012)*
- [LDOW'12] Birte Glimm, Aidan Hogan, Markus Krötzsch, Axel Polleres: OWL: Yet to arrive on the Web of Data? *In Linked Data on the Web (LDOW2012)*
- [JWS'11] Aidan Hogan, Andreas Harth, Jürgen Umbrich, Sheila Kinsella, Axel Polleres, Stefan Decker: Searching and browsing Linked Data with SWSE: The Semantic Web Search Engine. *JWS*. 9(4): 365-401 (2011)
- [LDOW'10] Aidan Hogan, Andreas Harth, Alexandre Passant, Stefan Decker, Axel Polleres: Weaving the Pedantic Web. LDOW 2010.
- [JWS'12a] Aidan Hogan, Jürgen Umbrich, Andreas Harth, Richard Cyganiak, Axel Polleres, Stefan Decker: An empirical survey of Linked Data conformance. *J. Web Sem.* 14: 14-44 (2012)
- [ESWC'13] Stefan Bischof, Axel Polleres: RDFS with Attribute Equations via SPARQL Rewriting. ESWC 2013: 335-350
- [RR'11] Renaud Delbru, Giovanni Tummarello, Axel Polleres: Context-Dependent OWL Reasoning in Sindice - Experiences and Lessons Learnt. RR 2011: 46-60
- [ISWC'10] Aidan Hogan, Jeff Z. Pan, Axel Polleres, Stefan Decker: SAOR: Template Rule Optimisations for Distributed Reasoning over 1 Billion Linked Data Triples. ISWC (1) 2010: 337-353
- [RR'09] Aidan Hogan, Stefan Decker: On the Ostensibly Silent 'W' in OWL 2 RL. RR 2009: 118-134
- [IJSWIS'09] Aidan Hogan, Andreas Harth, Axel Polleres: Scalable Authoritative OWL Reasoning for the Web. *Int. Journal On Semantic Web & Information Systems* 5(2): 49-90 (2009)
- ...(more on www.polleres.net/publications.html)

Backup-slides

Open Data Trends & Future



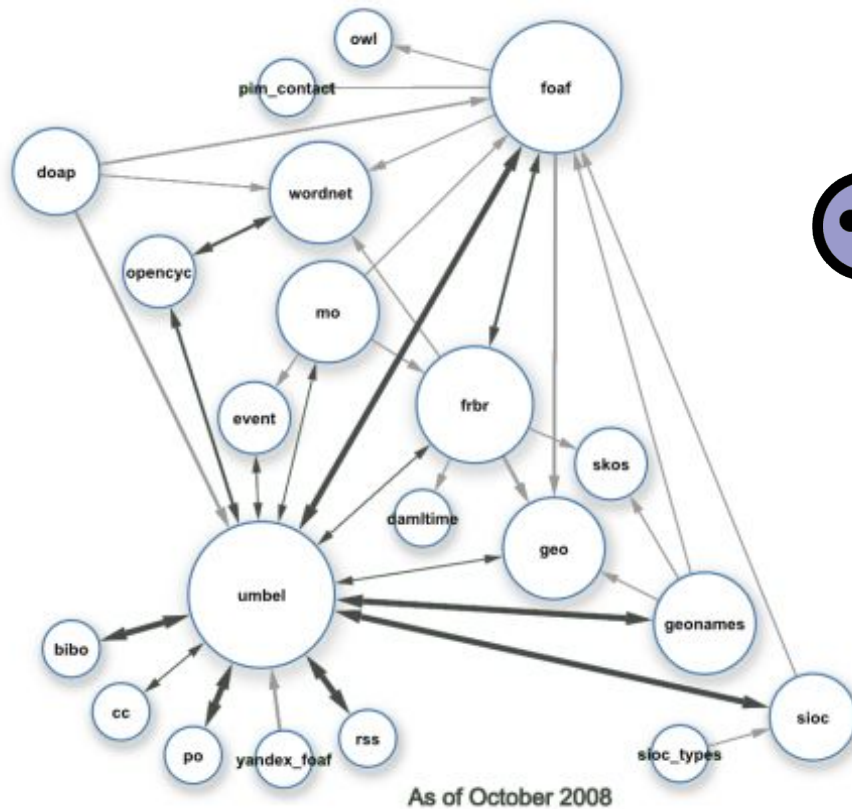
46

- Open Data: Typically very liberal licenses (variants of CC)
- Many formats, varying quality, harmonization starting
- Mostly by online communities or public bodies (cities, communities, governments, UN,...)
 - Currently focused mostly in SMEs to take advantage of that data
- vs. Publicly available data: e.g. NYT is public but not free/not license free
- vs. Enterprise (Linked) Data

Linked Data Ontologies = RDF Vocabularies (OWL, RDFS)



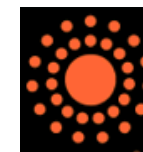
DOAP



schema.org



FOAF



DublinCore



SKOS



Image from http://blog.dbtune.org/public/081005_lod_constellation_m.jpg; Giasson, Bergman

*Side remark: Only a small fraction of Linked data is OWL/RDFS ontologies
(less than ~0.1% [IJSWIS'2009, Hogan 2010])*

OWL Reasoning: Complexity

- <http://www.w3.org/TR/2009/REC-owl2-profiles-20091027/>

Language	Reasoning Problems	Taxonomic Complexity	Data Complexity	Query Complexity	Combined Complexity
W3C Recommendation OWL 2 RDF-Based Semantics	Ontology Consistency, Class Expression Satisfiability, Class Expression Subsumption, Instance Checking, Conjunctive Query Answering	Undecidable	Undecidable	Undecidable	Undecidable
	Ontology Consistency, Class Expression Satisfiability, Class Expression Subsumption, Instance Checking	2NEXPTIME-complete (NEXPTIME if property hierarchies are bounded)	Decidable, but complexity open (NP-Hard)	Not Applicable	2NEXPTIME-complete (NEXPTIME if property hierarchies are bounded)
	Conjunctive Query Answering	Decidability open	Decidability open	Decidability open	Decidability open

And even for OWL RL PTIME-complete.... i.e., often too expensive.

Related Works:

• Scalable OWL Reasoning

- Jesse Weaver, James A. Hendler: Parallel Materialization of the Finite RDFS Closure for Hundreds of Millions of Triples. International Semantic Web Conference 2009: 682-697
- Vladimir Kolovski, Zhe Wu, George Eadon: Optimizing Enterprise-Scale OWL 2 RL Reasoning in a Relational Database System. International Semantic Web Conference (1) 2010: 436-452
- Jacopo Urbani, Spyros Kotoulas, Jason Maassen, Frank van Harmelen, Henri E. Bal: WebPIE: A Web-scale Parallel Inference Engine using MapReduce. J. Web Sem. 10: 59-75 (2012)

RDFS only, syntactic data. Nice hardware.

Similar ideas to ours in terms of rule optimization by ORACLE (parallel to our results)

Similar parallelization techniques to ours, 100b triples, but confined data with small ontologies. No Web Data. Nice: exchange of results for joins in Map-reduce jobs.

• Live Linked Data Querying

- Olaf Hartig, Christian Bizer, Johann Christoph Freytag: Executing SPARQL Queries over the Web of Linked Data. International Semantic Web Conference 2009: 293-309
- Günter Ladwig, Thanh Tran: Linked Data Query Processing Strategies International Semantic Web Conference (1) 2010: 453-469

No reasoning considered. 4 fixed queries, result throughput not considered

Confined Data, no reasoning considered. 8 fixed queries, needs a source index