



# Semantics $\sqcap$ Scalability $\models \perp$ ?

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# The Semantic Web

- Web “invented” by **Tim Berners-Lee** (an Oxford graduate!), then a physicist working at CERN
- His original vision of the Web was much more **ambitious** than the reality of the existing (syntactic) Web:



“... a set of **connected applications** ... forming a **consistent logical web of data** ... information is given **well-defined meaning**, better enabling computers and people to work in cooperation ...”

- This vision of the Web has become known as the **Semantic Web**
- Latest (refined) definition:  
"a web of data that can be processed directly and indirectly by machines"

# Semantic Technologies

- Initial focus was on necessary **underpinning**, including:

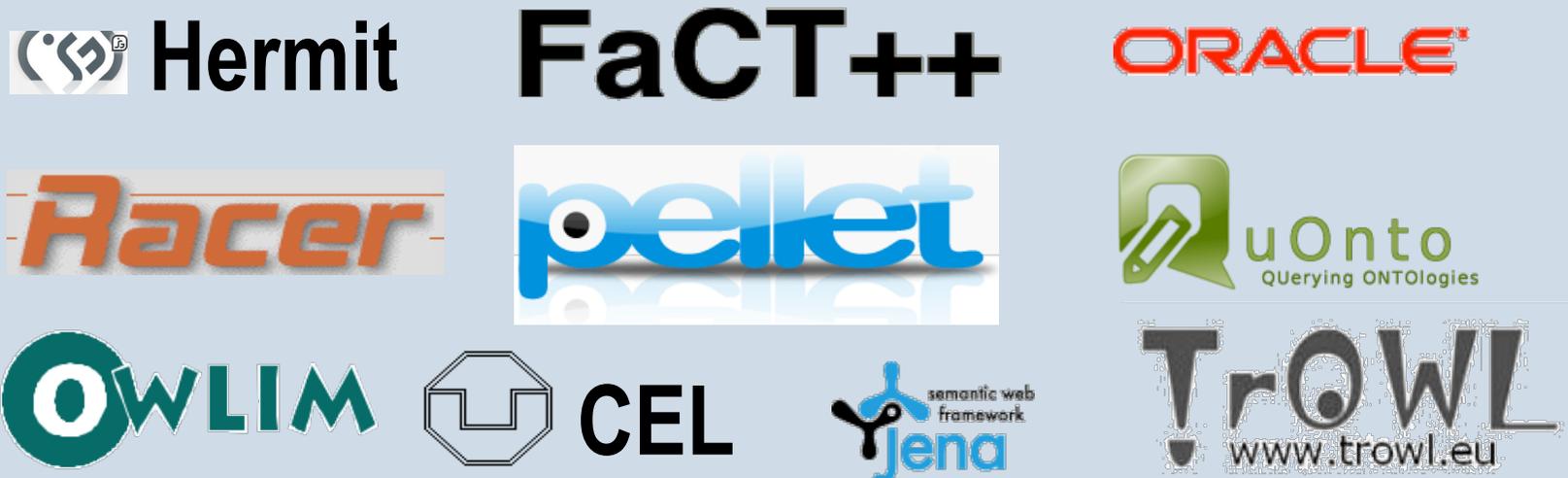
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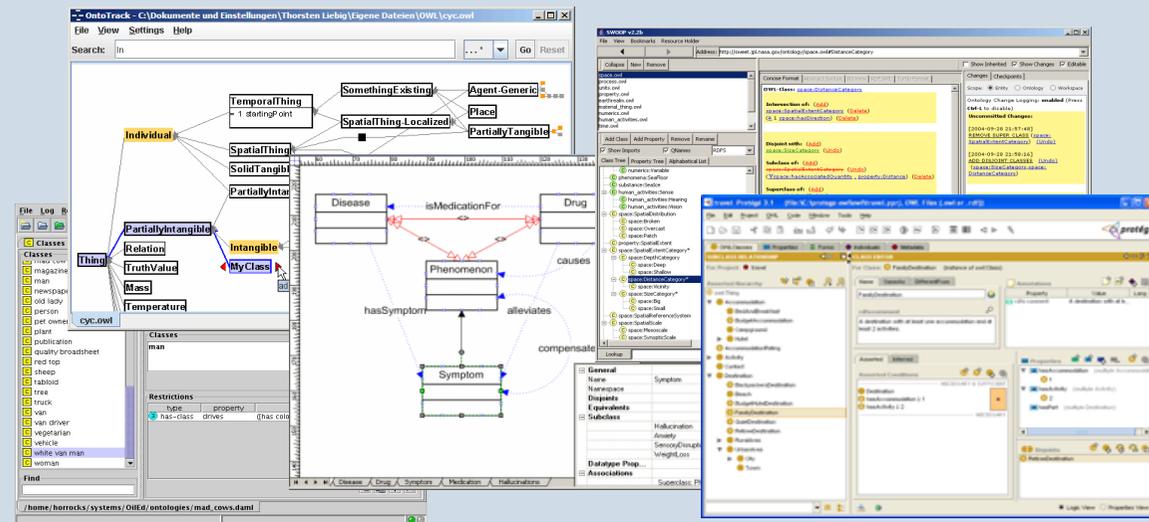
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- Storage and querying
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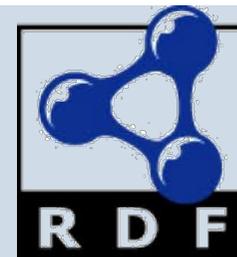


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- Resulting **robust infrastructure** used in SW applications

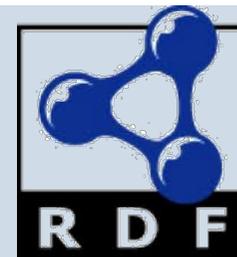
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  - Storage and querying
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- Resulting **robust infrastructure** used in SW applications
- Also increasingly used in “**Intelligent Information System**” applications



# How Does it Work?

- 1 **Standardised language for exchanging data**
  - W3C standard for data exchange is **RDF**
  - RDF is a simple language consisting of **<S P O> triples**
    - for example **<eg:Ian eg:worksAt eg:Oxford>**
    - all S,P,O are URIs or literals (data values)
  - **URIs** provides a flexible **naming scheme**
  - Set of triples can be viewed as a **graph**



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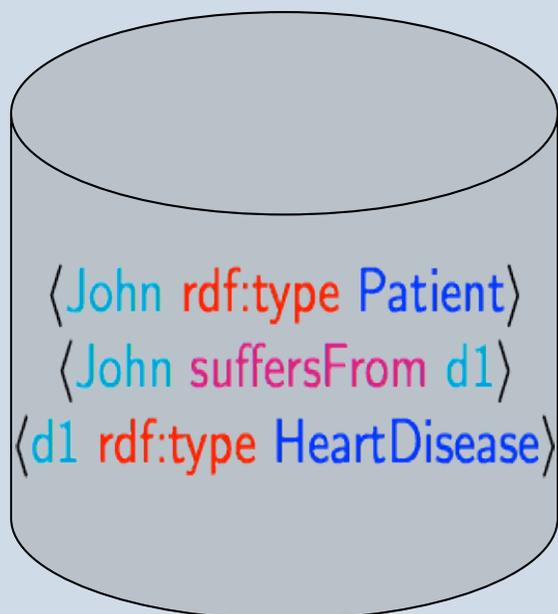
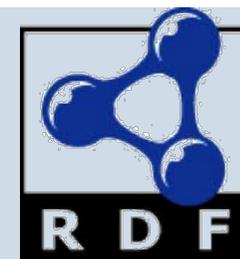
- 2 Standardised language for exchanging **vocabularies/schemas**
  - W3C standard for vocabulary/schema exchange is **OWL**
  - OWL provides for rich conceptual schemas, aka **ONTOLOGIES**

Heart  $\sqsubseteq$  MuscularOrgan  $\sqcap$   
 $\exists$ isPartOf.CirculatorySystem

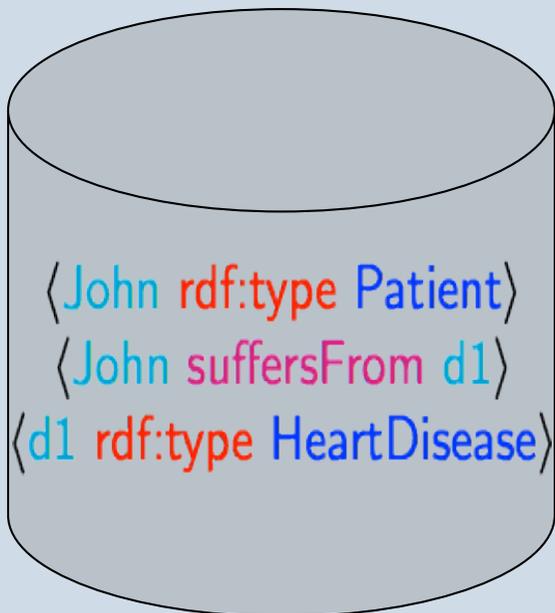
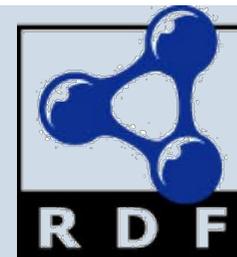
HeartDisease  $\equiv$  Disease  $\sqcap$   
 $\exists$ affects.Heart

VascularDisease  $\equiv$  Disease  $\sqcap$   
 $\exists$ affects.( $\exists$ isPartOf.CirculatorySystem)

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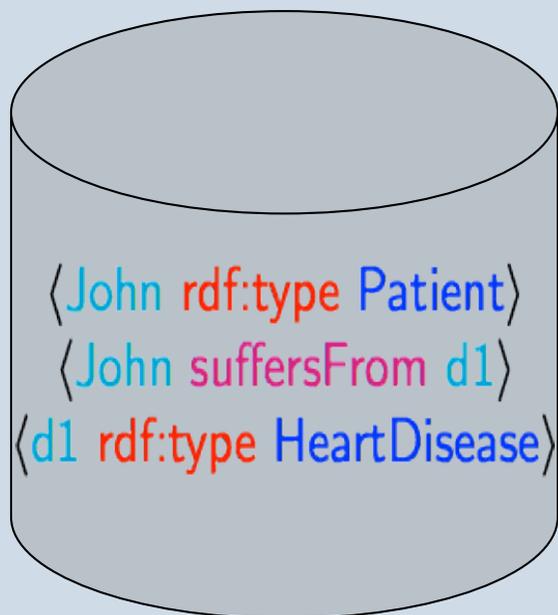
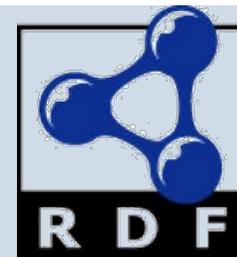
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Patients suffering from vascular disease



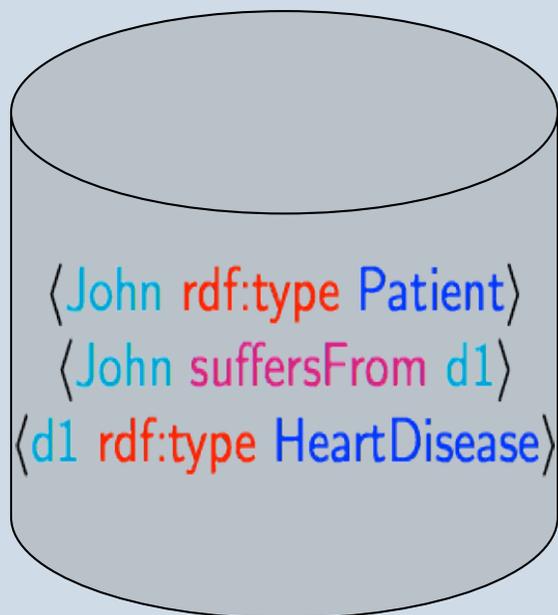
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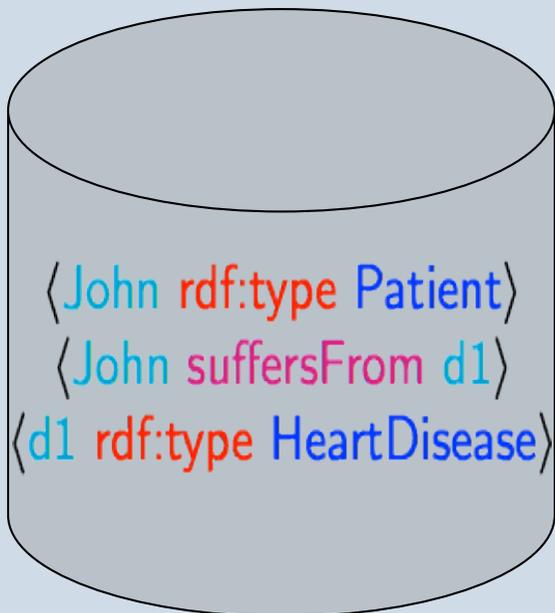
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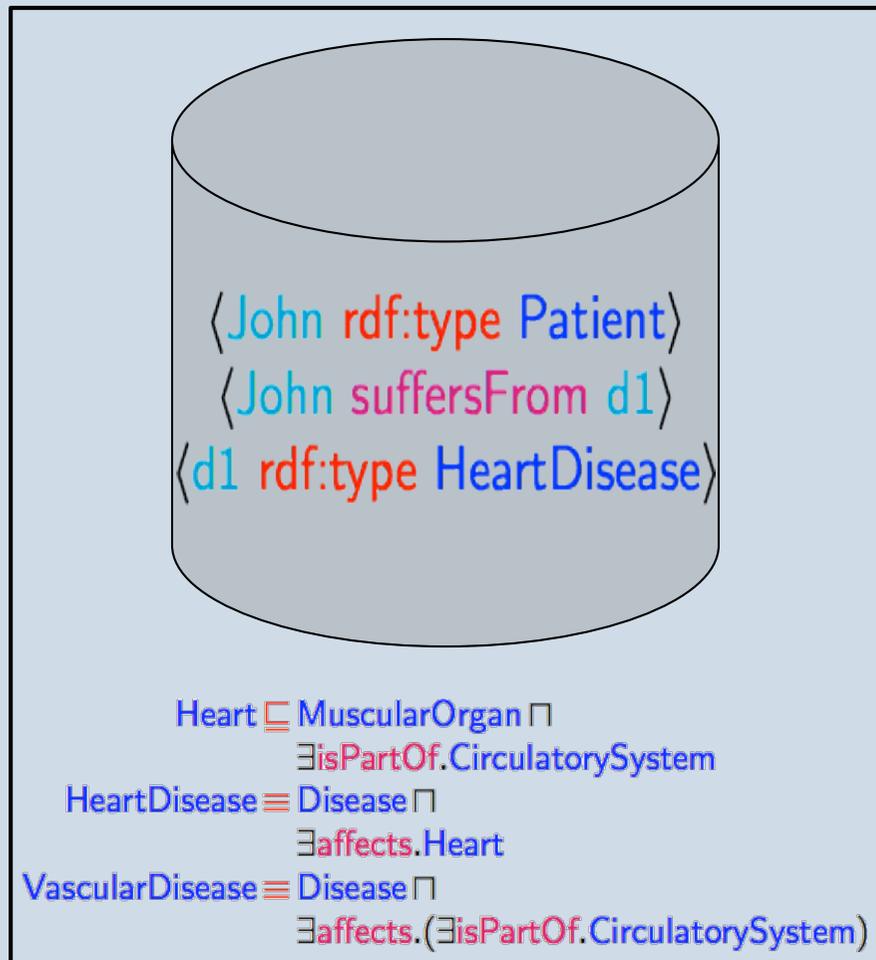
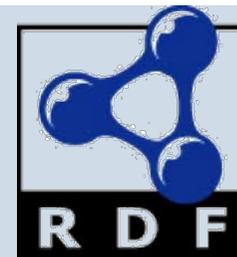


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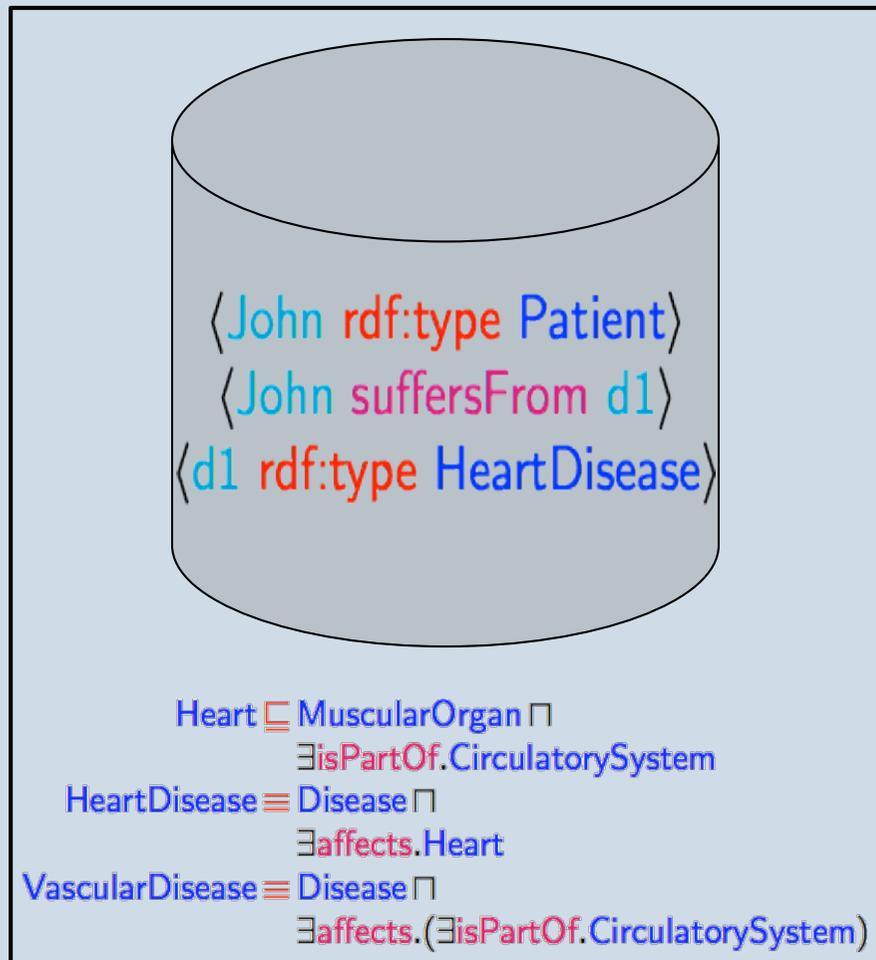
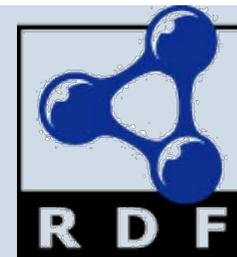
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Patients suffering from vascular disease

John



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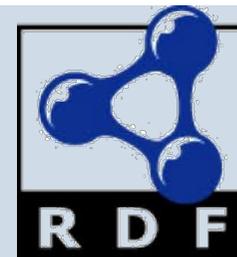


Diagram illustrating a database containing RDF triples and a set of class definitions:

Database contents:

- `<John rdf:type Patient>`
- `<John suffersFrom d1>`
- `<d1 rdf:type HeartDisease>`

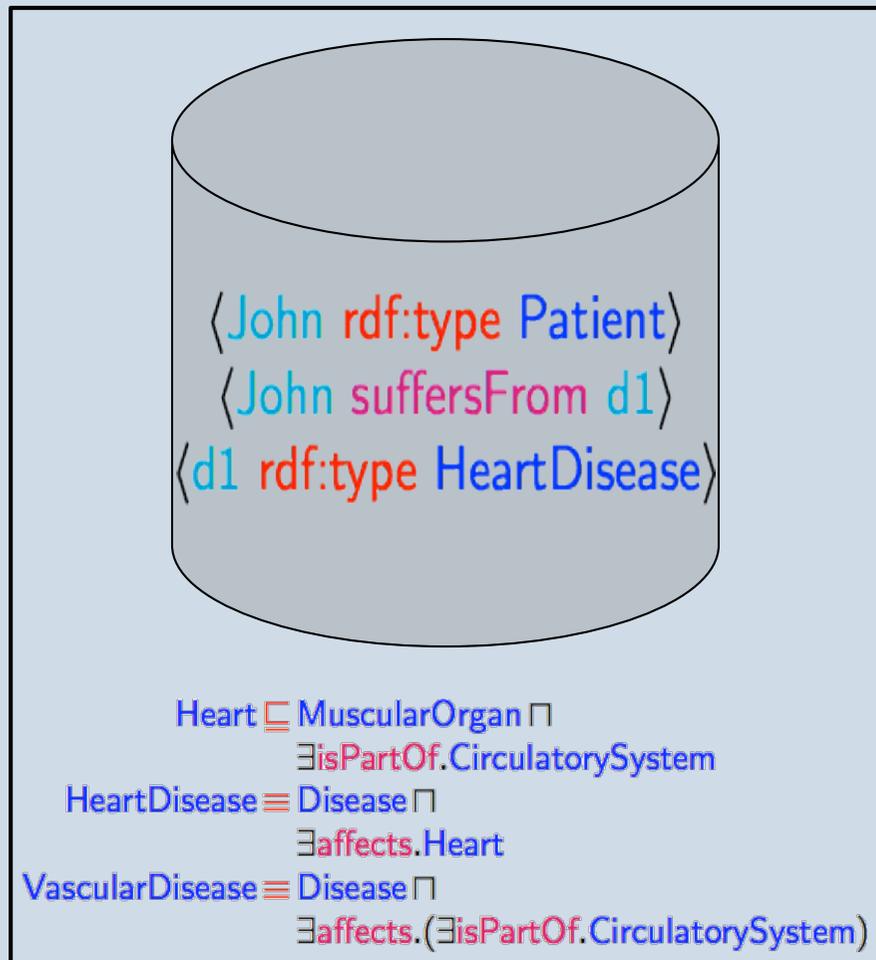
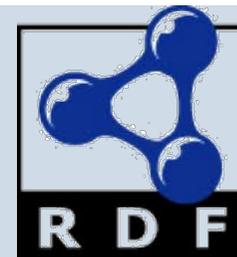
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Is heart disease a kind of vascular disease?



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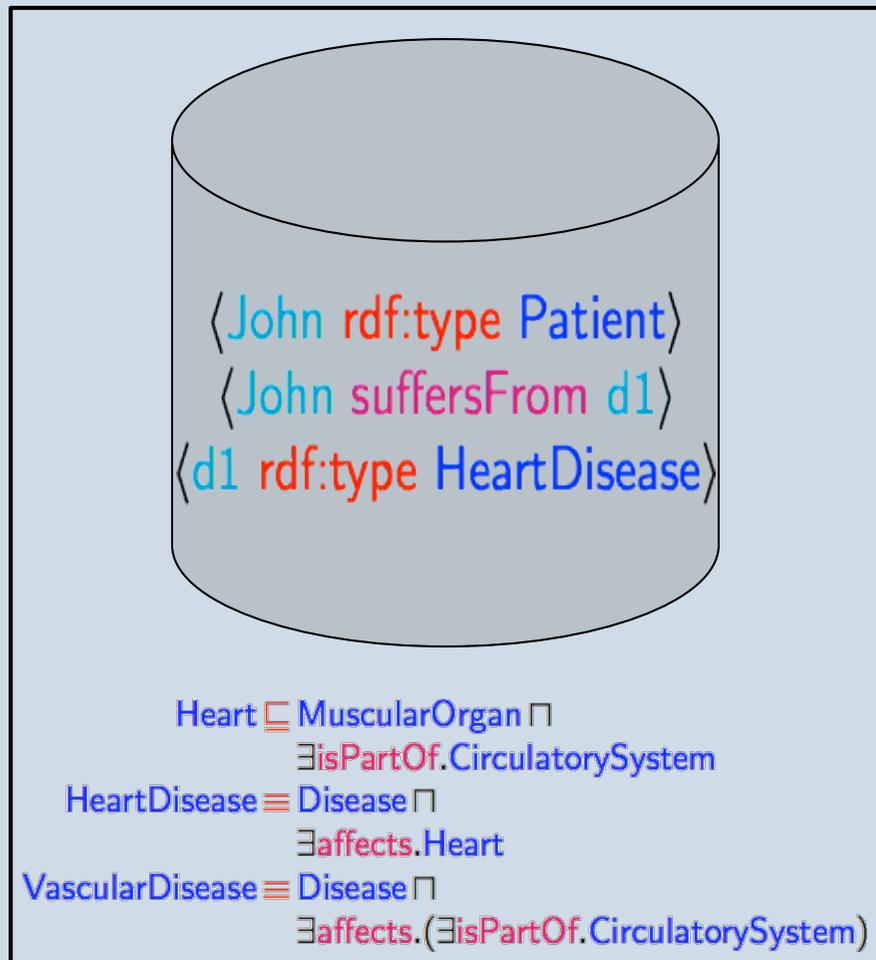
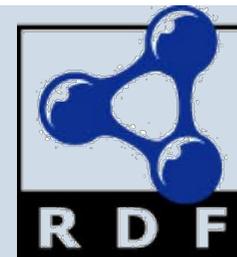


Is heart disease a kind of vascular disease?

YES



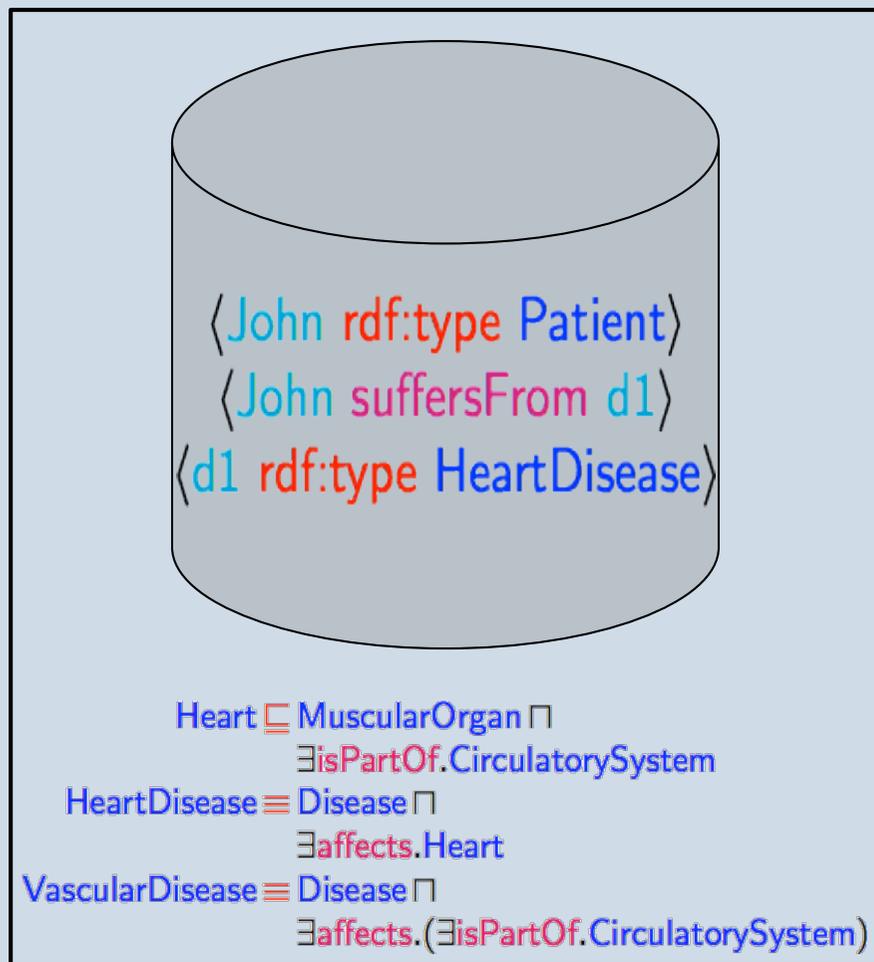
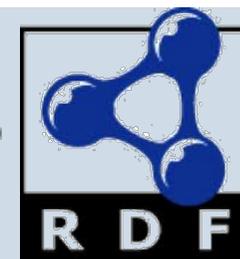
# How Does it Work?



Why?



# How Does it Work?



Why?

Heart  $\Rightarrow$   $\exists$ isPartOf.CirculatorySystem, ...



# Applications: Semantic Web

Text only | Help

**BBC** Home News Sport Weather iPlayer TV Radio More... Search

**SPORT WORLD CUP 2010**

SPORT FOOTBALL WORLD CUP 2010 GROUPS & TEAMS FIXTURES & RESULTS VIDEO BBC COVERAGE

Latest matches

**NED 2-1 BRA**  
  
[Highlights & report](#)

**URU 1-1 GHA**  
  
[Highlights & report](#)

**ARG 0-4 GER**  
  
[Highlights & report](#)

**PAR 0-1 ESP**  


**England**

[England 1-1 United States](#) Saturday, 12 June [Match report](#)  
[England 0-0 Algeria](#) Friday, 18 June [Match report](#)  
[Slovenia 0-1 England](#) Wednesday, 23 June [Match report](#)  
[Germany 4-1 England](#) Sunday, 27 June [Match report](#)

	A	B	C	D	E	F	G	H
<b>Group C Teams</b>								
USA				W	D	L	GD	PTS
England				1	2	0	1	5
Slovenia				1	1	1	0	4
Algeria				0	1	2	-2	1

**Latest stories**

 **Gerrard commits future to England** NEW  

- England sponsorship likely to end
- Capello to remain England manager
- Mueller blames England imbalance
- Capello receives Gartside backing

 **Pressure got to Rooney - Ferguson**  

- FA unfit for purpose says Caborn
- England's fear of crossing borders
- England duo bypass London event
- Barwick baffled by dismal England

**Features**

 **German lessons**  
 Jurgen Klinsmann on how to revolutionise England

- A German view on English football
- Redknapp backs England to shine
- BBC pundits on England
- Roy Hodgson Q&A
- World Cup goals analysis

**Around the web**

- BBC Search+ country page
- England Fifa Profile



# Applications: Semantic Web



# Applications: HCLS

- **SNOMED-CT** (Clinical Terms) ontology
  - provides common vocabulary for recording clinical data
  - used in healthcare systems of more than 15 countries, including Australia, Canada, Denmark, Spain, Sweden and the UK
  - “classified and checked for equivalencies” using ontology reasoners
- **OBO foundry** includes more than 100 biological and biomedical ontologies
  - “continuous integration server running **Elk and/or HermiT** 24/7 checking that multiple independently developed ontologies are mutually consistent”
- **Siemens** “actively building OWL based clinical solutions”

# Applications: Energy Supply Industry

- **EDF Energy** offer personalised energy saving advice to every customer
- **OWL ontology** used to model relevant environmental factors
- **HerMiT reasoner** used to match customer circumstances with relevant pieces of advice



# Applications: Intelligent Mobile Platform

- **Samsung** developing Intelligent Mobile Platform to support context-aware applications
- IMP monitors environment via **sensor data** (GPS, compass, accelerometer, ...)
- **OWL ontology** used to model environment and **infer context** (e.g., coffee with friends)
- Applications exploit context to enable more **intelligent behaviour**



# Applications: Oil and Gas Industry

- **Statoil** use data to inform production and exploration management
  - Large and complex data sets are difficult and time consuming to use
- Semantic technology can **improve access** to relevant data
- Test deployment in EU project **Optique**



# Theory $\rightsquigarrow$ Practice

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- OWL based on **description logic** *SROIQ*



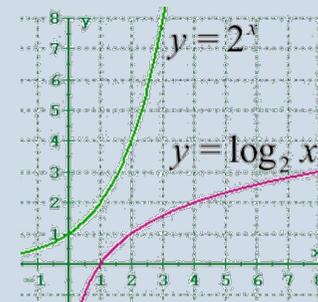
# Theory $\rightsquigarrow$ Practice

- OWL based on **description logic *SROIQ***
- DLs are a family of **FOL fragments**
  - Clear semantics
  - Well understood computational properties (e.g., decidability, complexity)
  - Simple goal directed reasoning algorithms



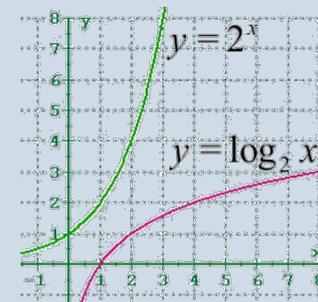
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  - NP-hard data complexity (-v- logspace for databases)



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**How can we provide robustly scalable query answering?**

# Various Approaches — Different Tradeoffs

① Use **full power of OWL** and a complete reasoner:

- ✓ Well-suited for modeling complex domains
- ✓ Reliable answers
- ✗ High worst-case complexity
- ✗ Scalability problems for large ontologies & datasets

## Complete OWL reasoners:

- E.g., FaCT++, **Hermit**, Pellet, ...
- Based on (hyper)tableau (model construction) theorem provers
- Highly optimised implementations effective on many ontologies, but not robust and unlikely to scale to large data sets

# Various Approaches — Different Tradeoffs

② Use a suitable “profile” and specialised reasoner:

**OWL 2** defines language subsets, aka **profiles** that can be “more simply and/or efficiently implemented”

- **OWL 2 EL**

- Based on  $\mathcal{EL}^{++}$
- PTime-complete for combined and data complexity

- **OWL 2 QL**

- Based on DL-Lite
- $AC^0$  data complexity (same as DBs)

- **OWL 2 RL**

- Based on “**Description Logic Programs**” ( $\approx DL \cap LP$ )
- PTime-complete for combined and data complexity

# Various Approaches — Different Tradeoffs

② Use a suitable “profile” and specialised reasoner:

- ✓ Tractable query answering
- ✓ Reliable answers (for inputs in the profile)
- ✗ Restricted expressivity of the ontology language
- ✗ Reasoners reject inputs outside profile

## OWL 2 EL ontology reasoners:

- E.g., CEL, ELK, ...
- Based on “consequence based” (deduction) theorem provers
- Target HCLS applications where many ontologies are (mainly) in the EL profile

# Schema Reasoning — Solved Problem?

	SNOMED CT	GALEN	FMA	GO
Logic	$\mathcal{EL}$	$\mathcal{EL}$	$\mathcal{EL}$	$\mathcal{EL}$
#classes	315,489	23,136	78,977	19,468
#properties	58	950	7	1
#axioms	430,844	36,547	121,712	28,897
# $\sqsubseteq$	$> 10^{11}$	$> 10^8$	$> 10^9$	$> 10^8$
ELK (1 worker)	13.15	1.33	0.44	0.20
ELK (4 workers)	5.02	0.77	0.39	0.19

	Plant Anat.	SWEET-P	NCI-2	DOLCE-P
Logic	$\mathcal{SHIF}$	$\mathcal{SHOIN}$	$\mathcal{ALCH}$	$\mathcal{SHOIN}$
#classes	19,145	1,728	70,576	118
#properties	82	145	189	264
#axioms	35,770	2,419	100,304	265
# $\sqsubseteq$	$> 10^8$	$> 10^6$	$> 10^9$	$> 10^4$
HermiT	11.2	11.2	—	105.1
Pellet	87.2	—	172.0	105.1
FaCT++	22.9	0.2	60.7	—

# Schema Reasoning — Solved Problem?

- Full expressive power may be needed to model, e.g.:
  - *non-viral pneumonia* (negation)
  - *infectious pneumonia* is caused by a *virus* or a *bacterium* (disjunction)
  - *double pneumonia* occurs in *two lungs* (cardinalities)
  - *groin* has a part that is part of the *abdomen*, and has a part that is part of the *leg* (inverse properties)
- Single non-EL axiom may incur massive performance penalty

# MORe Modular Reasoner

- Integrates powerful (slower) and weaker (faster) reasoners
- Exploits **module extraction** techniques to identify subset of ontology that can be completely classified using fast reasoner.
- Slower reasoner performs **as few computations as possible**
- Bulk of **computation delegated to faster reasoner**
- Current prototype integrates **Hermit** and **ELK** [1]

[1] Armas Romero, Cuenca Grau, and Horrocks. Modular Combination of Reasoners for Ontology Classification. In Proc. of ISWC 2012 (to appear).

# MORe Modular Reasoner

Ontology	$ \mathcal{O} \setminus \mathcal{O}_{\mathcal{L}} $	$ \Sigma^{\mathcal{L}} $	$ \mathcal{M}_{[\mathcal{O}, \overline{\Sigma^{\mathcal{L}}}]}$	Classif. time (seconds)			
				HermiT	MORe		
					total	HermiT	ELK
GO	0	100%	0%	7.1	2.2 (↓69.0%)	0	0.1
Gazeteer	0	100%	0%	838.1	28.2 (↓96.6%)	0	15.6
NCI	65	94.9%	15.4%	84.1	28.6 (↓66.0%)	15.8	3.3
Protein	12	98.1%	6.6%	11.4	2.9 (↓74.6%)	0.4	0.9
Biomodels	22,079	45.2%	66.4%	741.4	575.6 (↓22.4%)	540.1	2.6
cellCycle	1	> 99.9%	< 0.1%	–	13.9 ( – )	<0.1	4.9
NCI+CHEBI	65	95.6%	10.3%	116.6	34.0 (↓70.8%)	16.3	4.1
NCI+GO	65	96.7%	10.4%	110.0	37.6 (↓65.8%)	17.6	3.2
NCI+Mouse	65	96.0%	13.3%	93.7	31.0 (↓66.9%)	16.6	2.6

# OWL 2 EL — Data Retrieval Queries?

- PTime potentially problematical for very large datasets

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- PTime potentially problematical for very **large datasets**
- Various approaches:
  - **Materialise taxonomy** and use DBMS (incomplete reasoning)
  - “**Combined approach**” using materialisation + OBDA [2]
  - **Datalog** engine with (some form of) query rewriting [3]
  - Highly **optimised ABox** reasoners [4]

[2] Kontchakov, Lutz, Toman, Wolter, Zakharyashev: The Combined Approach to Ontology-Based Data Access. IJCAI 2011.

[3] Stefanoni, Motik, Horrocks: Small Datalog Query Rewritings for EL. DL 2012

[4] Kazakov, Kroetzsch, Simancik: Practical Reasoning with Nominals in the EL Family of Description Logics. KR 2012

# Various Approaches — Different Tradeoffs

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## OWL 2 QL ontology reasoners:

- E.g., QuOnto, **Requiem**, ...
- Based on query rewriting technique — ontology used to rewrite (expand) query
- Targets applications where data stored in RDBMS — aka **Ontology Based Data Access** (OBDA)

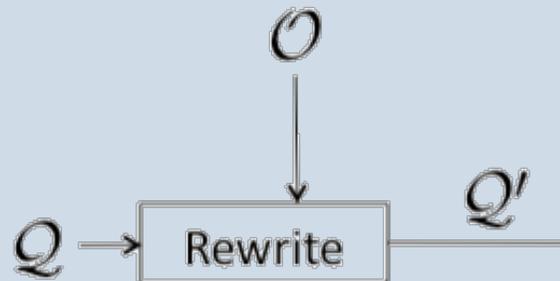
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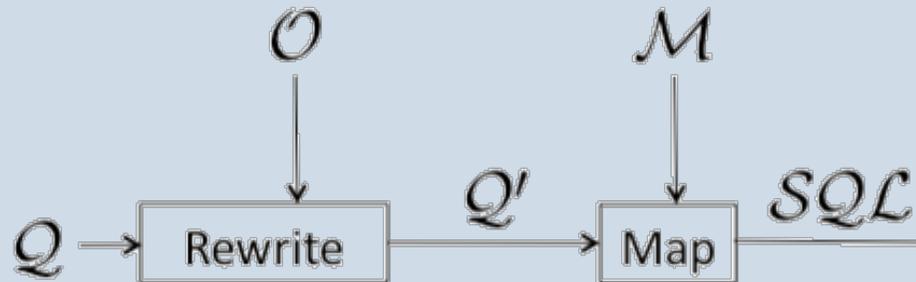
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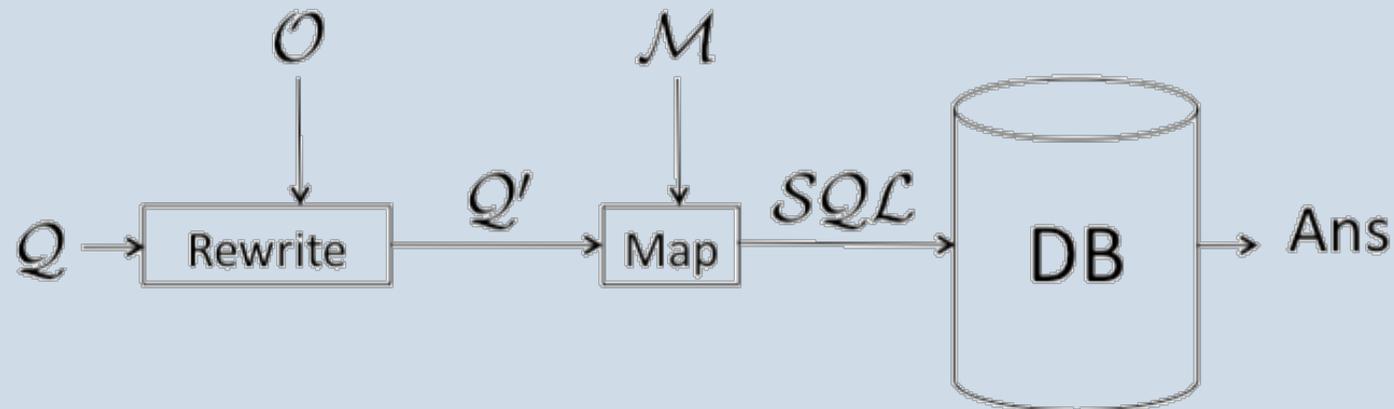
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- **Map** ontology queries  $\rightarrow$  DB queries (typically SQL) using mappings  $\mathcal{M}$  to rewrite  $Q'$  into a DB query
- **Evaluate** (SQL) query against DB



# Query Rewriting — Example

$$\mathcal{O} \left\{ \begin{array}{l} \text{Doctor} \sqsubseteq \exists \text{treats.Patient} \\ \text{Consultant} \sqsubseteq \text{Doctor} \end{array} \right.$$

$$\mathcal{Q} \quad Q(x) \leftarrow \text{treats}(x, y) \wedge \text{Patient}(y)$$

$$\mathcal{M} \left\{ \begin{array}{ll} \text{Doctor} & \mapsto \text{SELECT Name FROM Doctor} \\ \text{Patient} & \mapsto \text{SELECT Name FROM Patient} \\ \text{treats} & \mapsto \text{SELECT DName, PName FROM Treats} \end{array} \right.$$

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 \end{array}$$

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# Query Rewriting — Example

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$SQL$   $\left\{ \begin{array}{l} \text{SELECT Name FROM Doctor UNION} \\ \text{SELECT DName FROM Treats, Patient WHERE PName=Name} \end{array} \right.$

# Query Rewriting — Issues

## ① Rewriting

- May be large (worst case exponential in size of ontology)
- Queries may be hard for existing DBMSs
- Ongoing work on OBDA optimisation techniques, e.g., [5]

## ② Mappings

- May be difficult to develop and maintain
- Little work in this area to date

[5] Rodriguez-Muro, Calvanese: High Performance Query Answering over DL-Lite Ontologies. KR 2012

# Various Approaches — Different Tradeoffs

③ Use **full power of OWL** and incomplete reasoner:

- ✓ Well-suited for modeling complex domains
- ✓ Favourable scalability properties
- ✓ Flexibility: no inputs rejected
- ✗ Incomplete answers (and degree of incompleteness not known)

## **OWL 2 RL ontology reasoners:**

- E.g., Oracle's Semantic Datastore, Sesame, Jena, OWLim, ...
- Based on RDF triple stores and chase-like materialisation
- Widely used in practice to reason with large datasets
- Complete (only) for RL ontologies and ground atomic queries

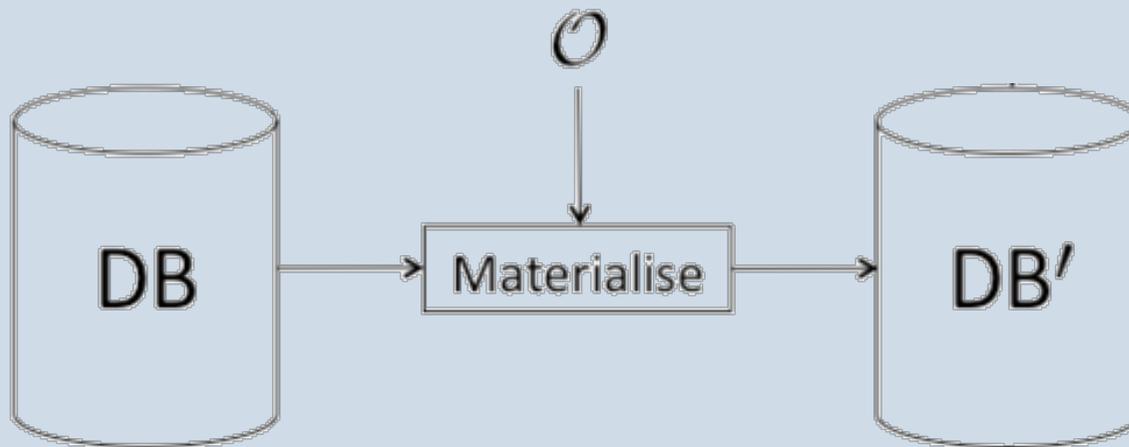
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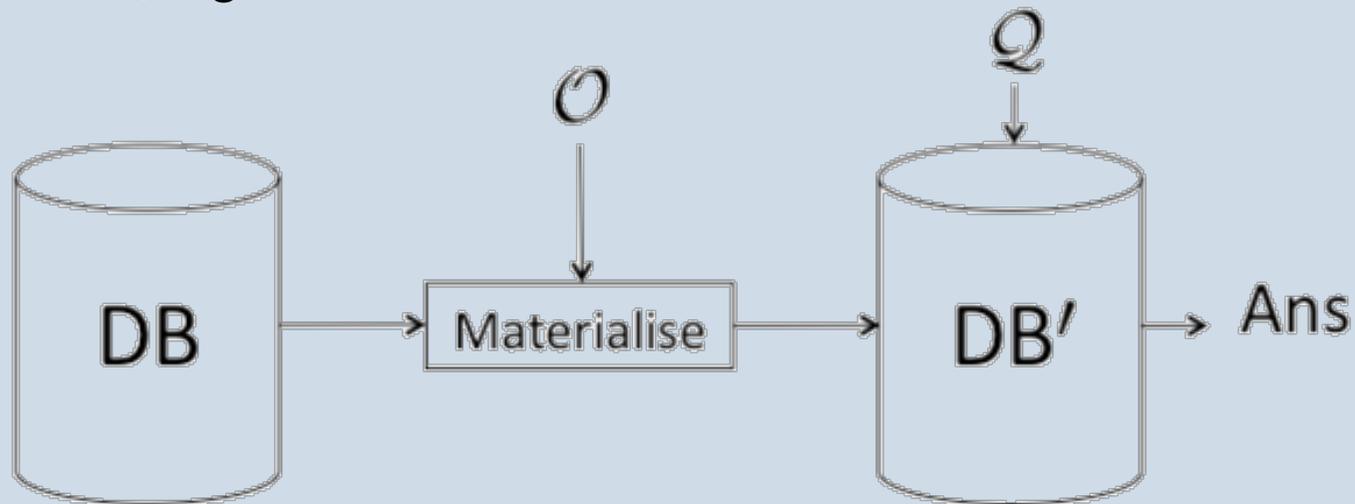
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  - nb: Closely related to **chase** procedure used with DB dependencies
- **Evaluate**  $Q$  against DB'



# Materialisation — Example

$\mathcal{O} \left\{ \begin{array}{l} \text{Doctor} \equiv \exists \text{treats.Patient} \\ \text{Consultant} \sqsubseteq \text{Doctor} \end{array} \right.$

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# Dealing With Frequently Changing Data

**Adding data** is relatively easy

- Monotonicity of FOL means that extending existing materialisation is sound
- Can still be quite costly if naively implemented

**Changing/retracting** data is much harder

- Naive solution requires all materialised facts to be discarded
- Re-materialisation very costly for large data sets
- But incremental reasoning is possible using view maintenance based techniques [6]

[6] Motik, Horrocks, and Kim. Delta-reasoner: a semantic web reasoner for an intelligent mobile platform. In Proc. of WWW 2012.

# Dealing with Incompleteness

- Materialisation based reasoning complete for **OWL 2 RL** profile (and ground atomic queries)
- But for ontologies **outside the profile**:
  - Reasoning may be incomplete
  - Incompleteness difficult to measure via empirical testing
- Possible solutions offered by recent work:
  - **Measuring and repairing incompleteness**
  - **Chase materialisation**
  - **Computing upper and lower bounds**

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  - $\mathbf{S}_Q$  a set of pairs  $\langle \mathcal{A}, \mathcal{Y} \rangle$  with  $\mathcal{A}$  an ABox and  $\mathcal{Y}$  a query
- A **reasoner**  $\mathcal{R}$  passes  $\mathbf{S}$  if:
  - $\mathcal{R}$  finds  $\mathcal{O} \cup \mathcal{A}$  unsatisfiable for each  $\mathcal{A} \in \mathbf{S}_\perp$
  - $\mathcal{R}$  complete for  $\mathcal{Y}$  w.r.t.  $\mathcal{O} \cup \mathcal{A}$  for each  $\langle \mathcal{A}, \mathcal{Y} \rangle \in \mathbf{S}_Q$

[7] Cuenca Grau, Motik, Stoilos, and Horrocks. Completeness Guarantees for Incomplete Ontology Reasoners: Theory and Practice. JAIR, 43:419-476, 2012.

# Chase Materialisation

- Applicable to **acyclic** ontologies
  - Acyclicity can be checked using, e.g., graph based techniques (weak acyclicity, **joint acyclicity**, etc.)
  - Many realistic ontologies turn out to be acyclic
- Given acyclic ontology  $\mathcal{O}$ , can apply chase materialisation:
  - Ontology translated into **existential rules** (aka dependencies)
  - Existential rules can introduce **fresh Skolem individuals**
  - **Termination guaranteed** for acyclic ontologies

[8] Cuenca Grau et al. Acyclicity Conditions and their Application to Query Answering in Description Logics. In Proc. of KR 2012.

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$\mathcal{O} \left\{ \begin{array}{l} \text{Doctor} \equiv \exists \text{treats.Patient} \\ \text{Consultant} \sqsubseteq \text{Doctor} \end{array} \right.$

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# Computing Lower and Upper Bounds

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- RL reasoning w.r.t. OWL ontology  $\mathcal{O}$  gives **lower bound** answer  $L$
- Transform  $\mathcal{O}$  into **strictly stronger OWL RL ontology**
  - Transform ontology into Datalog<sup>±,∨</sup> rules
  - Eliminate  $\forall$  by transforming to  $\wedge$
  - Eliminate existentials by replacing with Skolem constants
  - Discard rules with empty heads
  - Transform rules into **OWL 2 RL ontology  $\mathcal{O}'$**

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- RL reasoning w.r.t.  $\mathcal{O}'$  gives (complete but unsound) upper bound answer  $U$
- If  $L = U$ , then both answers are sound and complete
- If  $L \neq U$ , then  $U \setminus L$  identifies a (small) set of “possible” answers
  - Indicates range of uncertainty
  - Can (more efficiently) check possible answers using, e.g., Hermit
  - Future work: use  $U \setminus L$  to identify (small) “relevant” subset of data needed to efficiently compute exact answer

[9] Zhou, Cuenca Grau, and Horrocks. Efficient Upper Bound Computation of Query Answers in Expressive Description Logics. In Proc. of DL 2012, volume 846 of CEUR.

# Discussion

Numerous **exciting developments** & research areas

- **Rewriting**: optimisations, extensions (datalog engines), etc.
- **Materialisation**: chase, repair, truth maintenance, upper bounds etc.
- **Combined** techniques (materialisation+rewriting), **Datalog**
- Specialised **RDF stores**, **Column** stores, massive **parallelism**, etc.
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Consider **progress on schema reasoning**:

Year	$O$ -size	Complete	Time (s)
1995	3,000	No	$10^5$
1998	3,000	Yes	300
2005	30,000	Yes	30
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**Looking forward to similar progress  
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## Semantics $\not\equiv$ Scalability $\perp$ !

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# Acknowledgements



Engineering and Physical Sciences  
Research Council



SIXTH FRAMEWORK  
PROGRAMME



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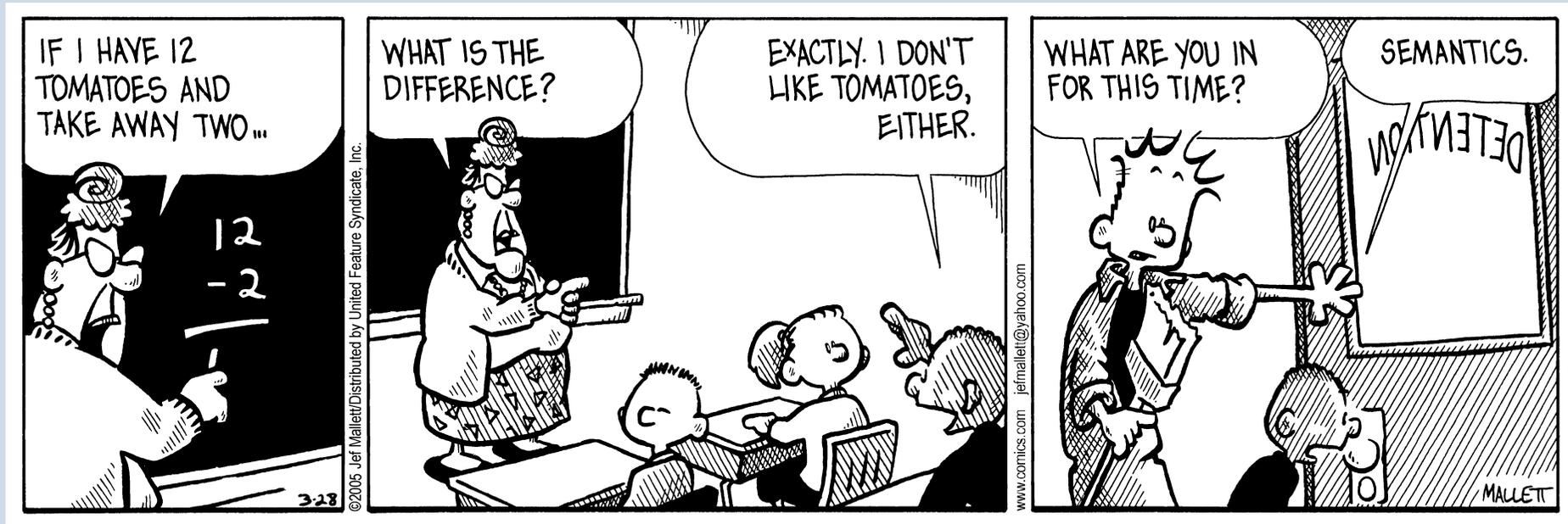


SEVENTH FRAMEWORK  
PROGRAMME

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# Thank you for listening



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# Any questions?