DEPARTMENT OF COMPUTER SCIENCE



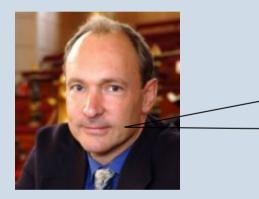


Semantic Technologies: Beyond the Semantic Web

Ian Horrocks Information Systems Group Department of Computer Science University of Oxford

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"









Initial focus was on necessary underpinning, including:









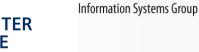
- Initial focus was on necessary underpinning, including:
 - Languages



Engineering and Physical Sciences

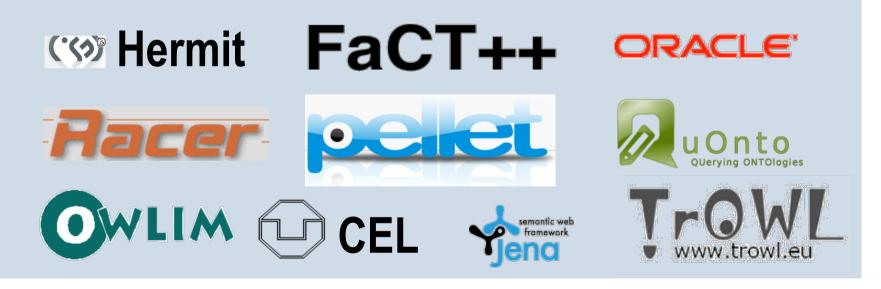
Research Council







- Initial focus was on necessary underpinning, including:
 - Languages
 - Storage and querying



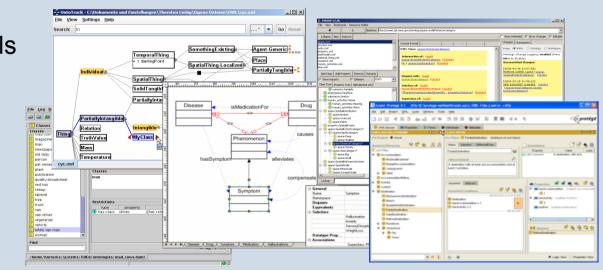








- Initial focus was on necessary underpinning, including:
 - Languages
 - Storage and querying
 - Development tools











- Initial focus was on necessary underpinning, including:
 - Languages
 - Storage and querying
 - Development tools
- Resulting robust infrastructure used in SW applications









- Initial focus was on necessary underpinning, including:
 - Languages
 - Storage and querying
 - Development tools
- Resulting robust infrastructure used in SW applications
- Also increasingly used in "Intelligent Information System" applications











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:lan 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









How Does it Work? Standardised language for exchanging data http://www.w3.org/2000/10/swap/pim/contact#Person eg:organisation http://www.w3.org/1999/02/22-rdf-syntax-ns#type http://...rdf-syntax-ns/#type http://www.w3.org/People/EM/contact#me eq:worksfor eg:w3c http://www.w3.org/2000/10/swap/pim/contact#fullName Eric Miller http://...fullName eg:hq http://www.w3.org/2000/10/swap/pim/contact#mailbox eg:Boston oles can be<mark>wac</mark>ewed a



Information Systems Group



Dr



mailto:em@w3.org

http://www.w3.org/2000/10/swap/pim/contact#personalTitle







1 Standardised language for exchanging data

	Triple		
	S	Р	0
• foi	em1234	rdf:type	Person
	em1234	name	"Eric Miller"
	em1234	title	"Dr"
	em1234	mailbox	mailto:em@w3.org
Set of	em1234	worksfor	w3c
	w3c	rdf:type	organisation
	w3c	hq	Boston
	w3c	name	"W3C"











Standardised language for exchanging data

C standard for data exchange is RDF							
PERSON							
ID	NAME	TITLE	MAILBOX	WORKSFOR			
em1234	"Eric Miller"	"Dr"	mailto:em@w3.org	y w3c			
ORGANISATION							
ID	NAME		HQ				
w3c	"W3C"		Boston				



1









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 $\exists affects.(\exists isPartOf.CirculatorySystem)$











B Standardised language for querying ontologies+data

- W3C standard for querying is SPARQL
- SPARQL provides a rich query language comparable to SQL
 - ?x worksfor ?y .
 ?y rdf:type organisation .
 ?y hq Boston .
 - Select ?x

where { ?x worksfor ?y . ?y rdf:type organisation . ?y hq Boston . }

Q(?x) ← worksfor(?x,?y) ∧ organisation(?y) ∧ hq(?y,Boston)









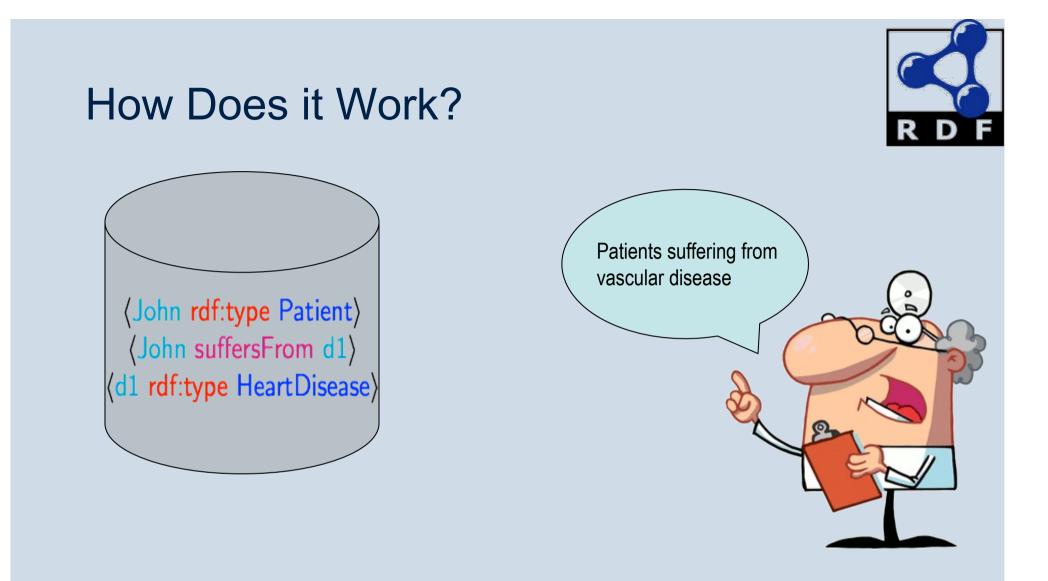










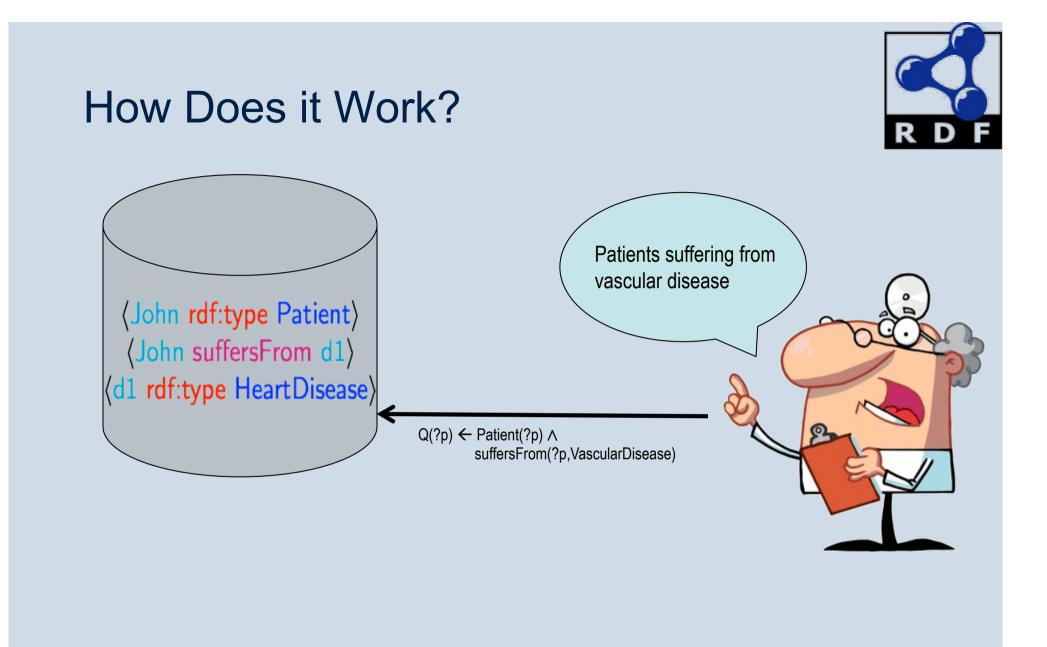










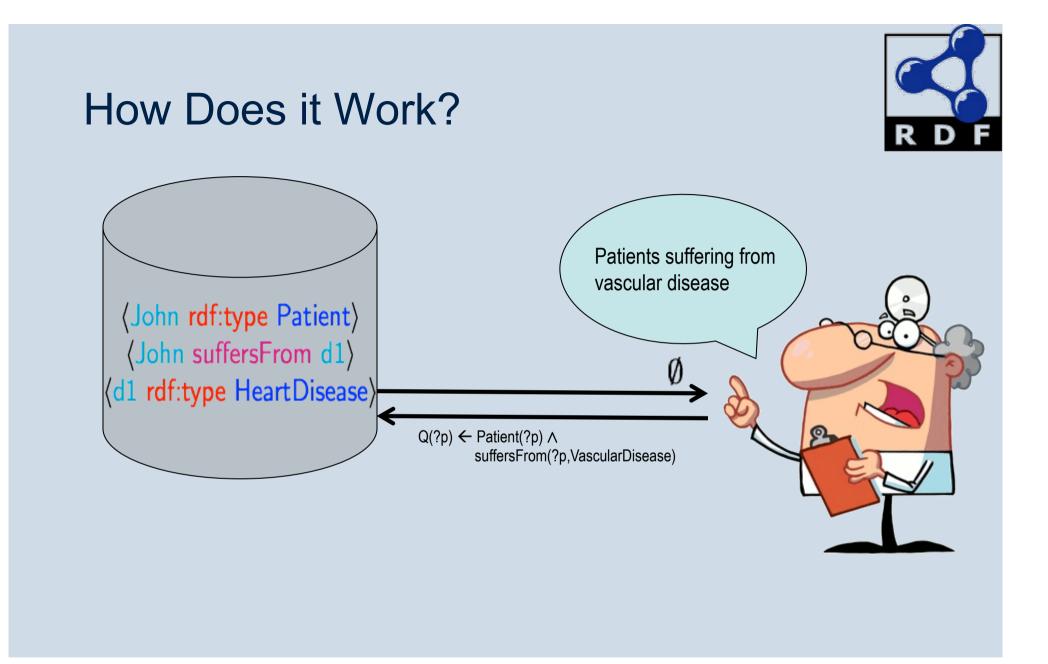










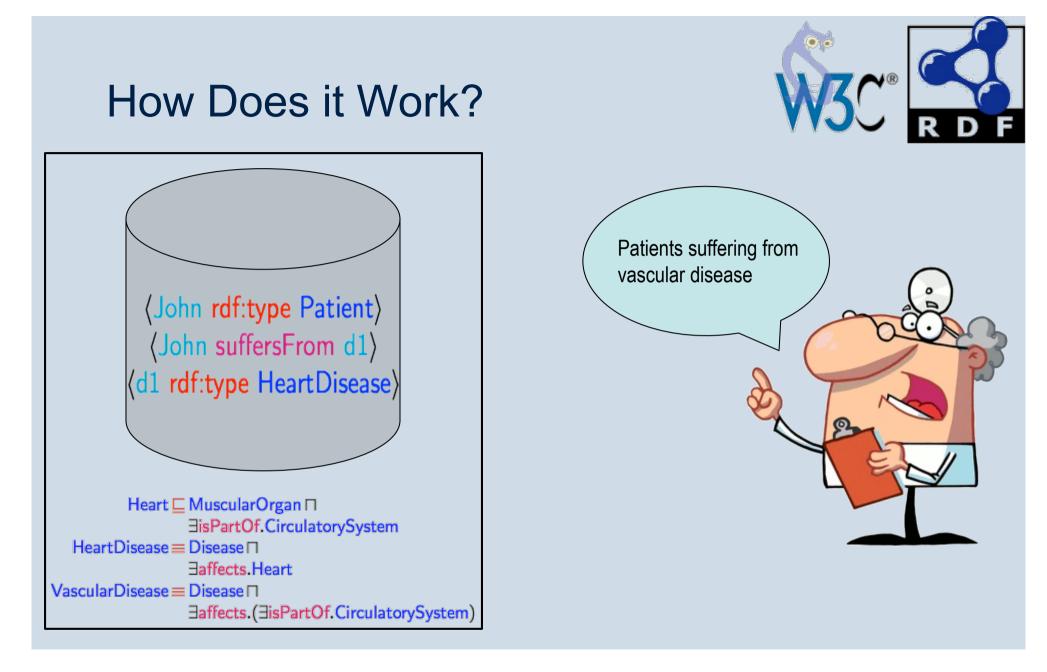










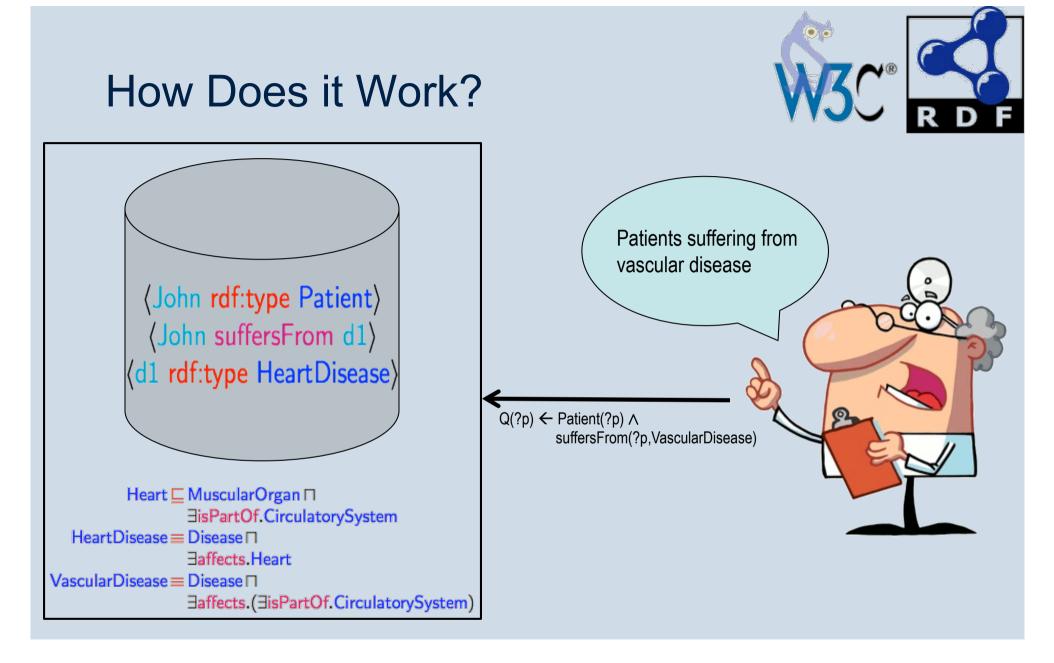










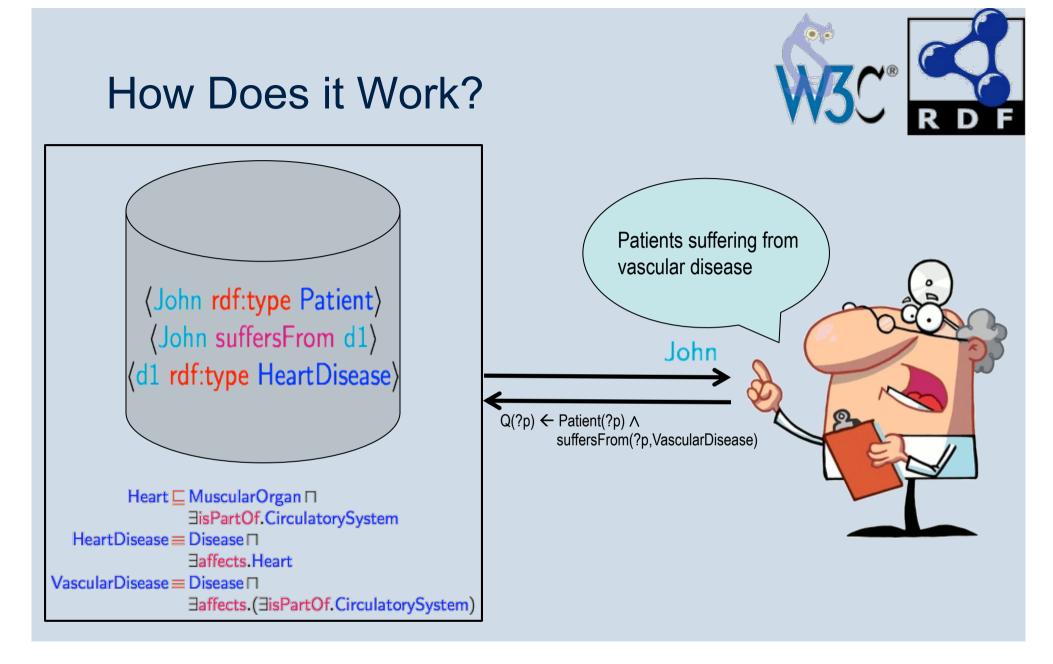










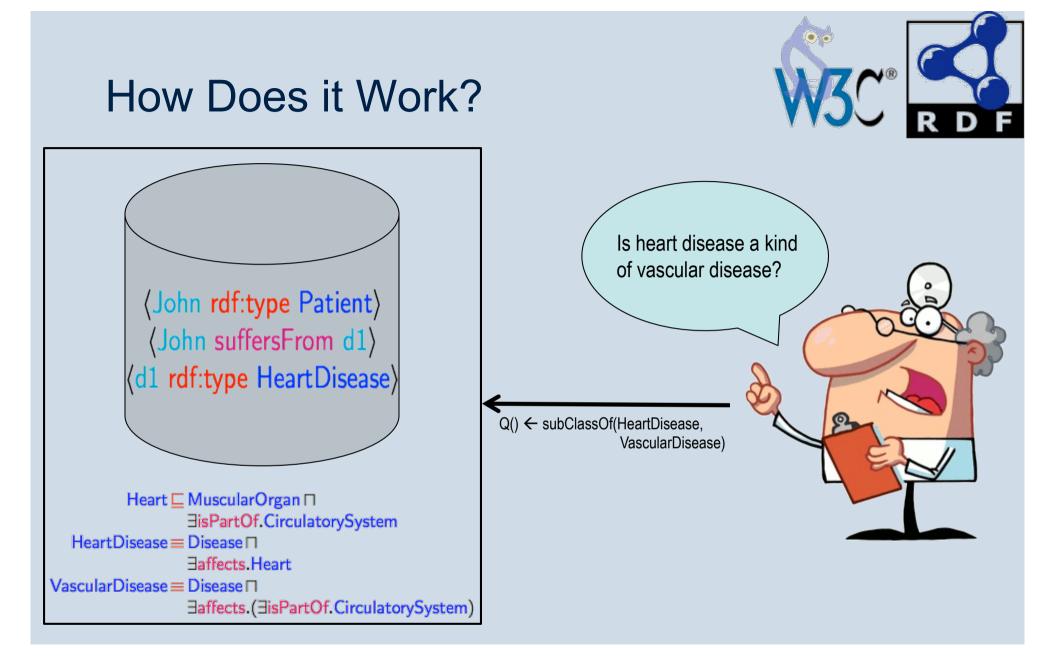










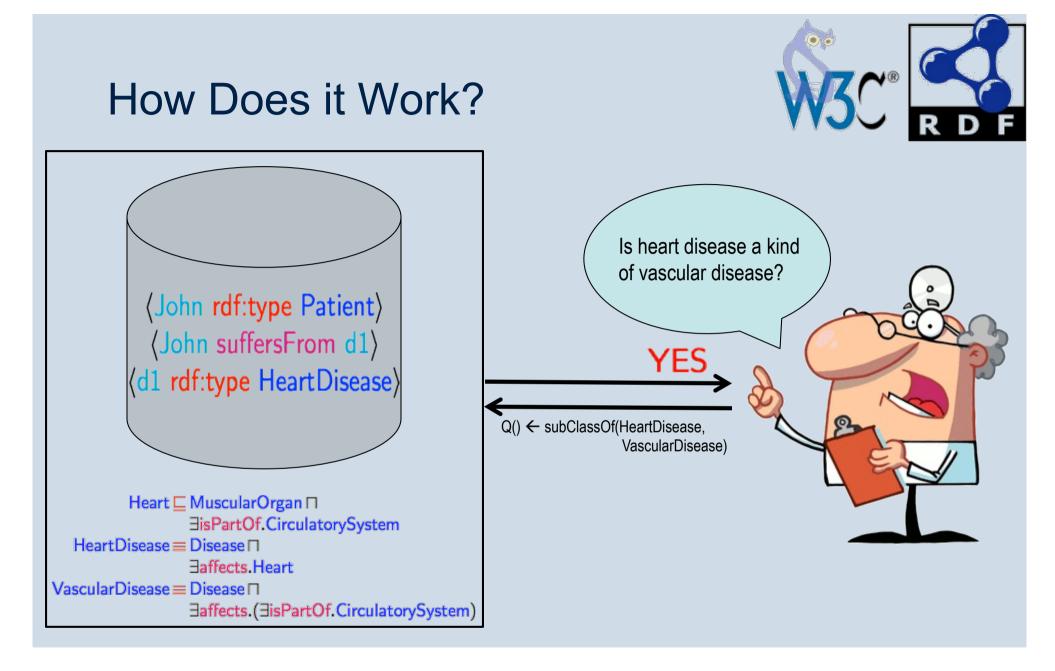










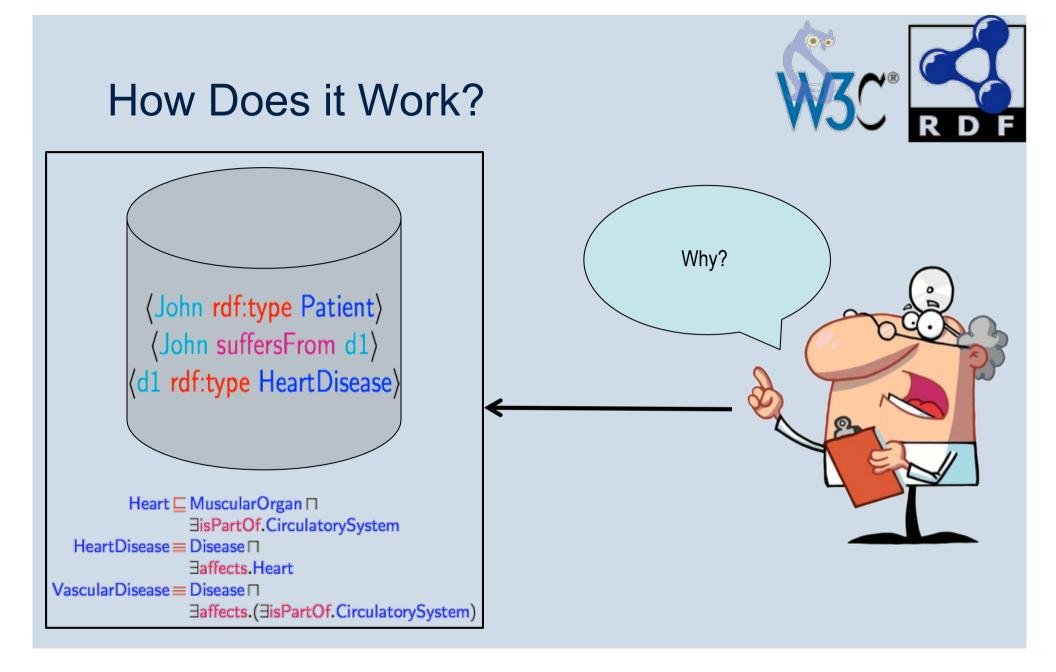










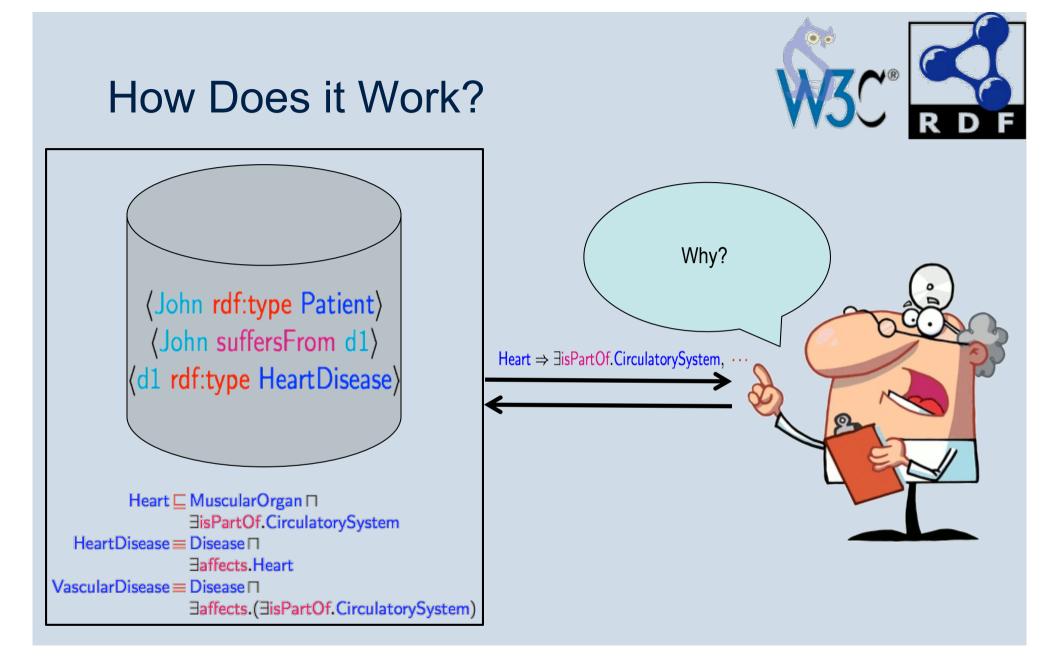












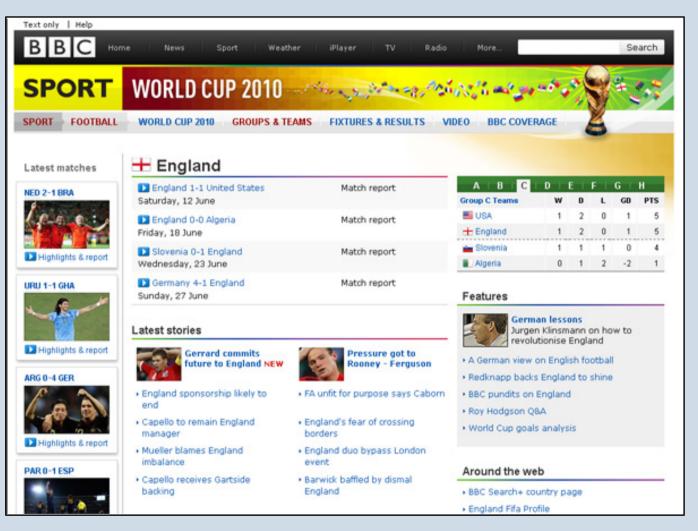








Applications: Semantic Web

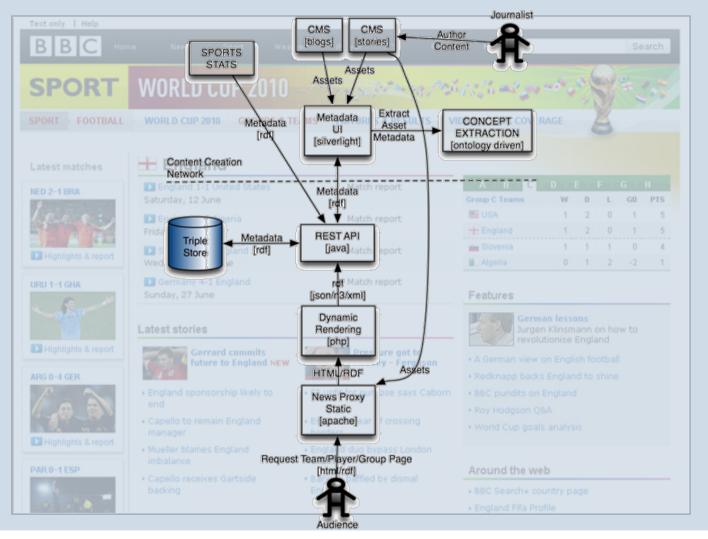








Applications: Semantic Web











Applications: Semantic Web











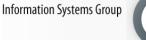
Applications: HCLS

- **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









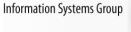


Applications: Intelligent Mobile Platform

- 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 ~~> Practice









Theory ~>> Practice

OWL based on description logic SROIQ









Theory ~>> 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







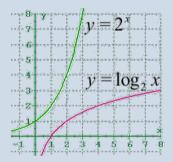




Theory ~>> Practice

- 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
- OWL is decidable, but highly highly intractable
 - N2ExpTime-comlete combined complexity
 - NP-hard data complexity (-v- logspace for databases)









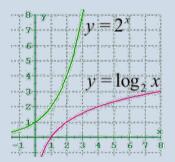




Theory ~> Practice

- 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
- OWL is decidable, but highly highly intractable
 - N2ExpTime-comlete combined complexity
 - NP-hard data complexity (-v- logspace for databases)





How can we provide robustly scalable query answering?









and now:

A Word from our Sponsors

















Decidable fragments of First Order Logic

Thank you for listening

Any questions?

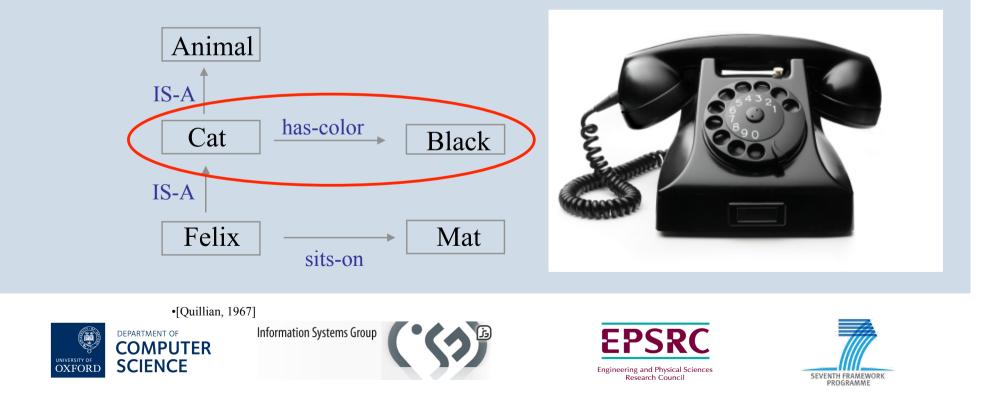








- A family of logic based Knowledge Representation formalisms
 - Originally descended from semantic networks and KL-ONE
 - Describe domain in terms of concepts (aka classes), roles (aka properties, relationships) and individuals



- Modern DLs (after Baader et al) distinguished by:
 - Fully fledged logics with formal semantics
 - Decidable fragments of FOL (often contained in C2)
 - Closely related to Propositional Modal/Dynamic Logics & Guarded Fragment
 - Computational properties well understood (worst case complexity)
 - Provision of inference services
 - Practical decision procedures (algorithms) for key problems (satisfiability, subsumption, query answering, etc)
 - Implemented systems (highly optimised)
- The basis for widely used ontology languages









- Signature
 - Concept (aka class) names, e.g., Cat, Animal, Doctor
 - Equivalent to FOL unary predicates
 - Role (aka property) names, e.g., sits-on, hasParent, loves
 - Equivalent to FOL binary predicates
 - Individual names, e.g., Felix, John, Mary, Boston, Italy
 - Equivalent to FOL constants







•

- Operators
 - Many kinds available, e.g.,
 - Standard FOL Boolean operators (□, □, ¬)
 - Restricted form of quantifiers (\exists, \forall)
 - Counting $(\geq, \leq, =)$









- Concept expressions, e.g.,
 - Doctor \sqcup Lawyer
 - Rich ⊓ Happy
 - Cat $\sqcap \exists sits \text{-on.Mat}$
- Equivalent to FOL formulae with one free variable
 - $Doctor(x) \lor Lawyer(x)$
 - $\operatorname{Rich}(x) \wedge \operatorname{Happy}(x)$
 - $Cat(x) \land \exists y.(sits-on(x, y))$









- Special concepts
 - \top (aka top, Thing, most general concept)
 - ⊥ (aka bottom, Nothing, inconsistent concept)

used as abbreviations for

- $(A \sqcup \neg A)$ for any concept A
- (A $\sqcap \neg$ A) for any concept A









- Role expressions, e.g.,
 - loves⁻
 - hasParent

 hasBrother
- Equivalent to FOL formulae with two free variables
 - loves(y, x)
 - $\exists z.(hasParent(x, z) \land hasBrother(z, y))$









- "Schema" Axioms, e.g.,
 - Rich ⊑ ¬Poor
 - Cat ⊓ ∃sits-on.Mat ⊑ Happy
 - BlackCat ≡ Cat ⊓ ∃hasColour.Black
 - sits-on ⊑ touches
 - Trans(part-of)

(concept inclusion)
(concept inclusion)
(concept equivalence)
(role inclusion)
(transitivity)

- Equivalent to (particular form of) FOL sentence, e.g.,
 - $\forall x.(\operatorname{Rich}(x) \rightarrow \neg \operatorname{Poor}(x))$
 - $\forall x.(Cat(x) \land \exists y.(sits-on(x,y) \land Mat(y)) \rightarrow Happy(x))$
 - $\forall x.(BlackCat(x) \leftrightarrow (Cat(x) \land \exists y.(hasColour(x,y) \land Black(y)))$
 - $\forall x, y.(sits-on(x, y) \rightarrow touches(x, y))$
 - $\forall x, y, z.((sits-on(x,y) \land sits-on(y,z)) \rightarrow sits-on(x,z))$











- "Data" Axioms (aka Assertions or Facts), e.g.,
 - BlackCat(Felix)
 - Mat(Mat1)
 - Sits-on(Felix,Mat1)
- Directly equivalent to FOL "ground facts"
 - Formulae with no variables

(concept assertion) (concept assertion) (role assertion)









• A set of axioms is called a TBox, e.g.:

{Doctor \sqsubseteq Person, Person \Box = Person \Box = heachild Person		
	Parent \equiv Person $\sqcap \exists$ hasChild.Person, HappyParent \equiv Parent $\sqcap \forall$ hasChild.(Do	Note
		Facts sometimes written
		John:HappyParent,
	{HappyParent(John),	John hasChild Mary,
	hasChild(John,Mary)}	(John,Mary):hasChild

- A Knowledge Base (KB) is just a TBox plus an Abox
 - Often written $\mathcal{K} = \langle \mathcal{T}, \mathcal{A} \rangle$









- Many different DLs, often with "strange" names
 - E.g., \mathcal{EL} , \mathcal{ALC} , \mathcal{SHIQ}
- Particular DL defined by:
 - Concept operators (□, □, ¬, ∃, ∀, etc.)
 - Role operators (⁻, o, etc.)
 - Concept axioms (\sqsubseteq , \equiv , etc.)
 - Role axioms (⊑, Trans, etc.)









- E.g., \mathcal{EL} is a well known "sub-Boolean" DL
 - Concept operators: □, ¬, ∃
 - No role operators (only atomic roles)
 - Concept axioms: <u></u>_, ≡
 - No role axioms
- E.g.:

Parent \equiv Person $\sqcap \exists$ hasChild.Person









- ALC is the smallest propositionally closed DL
 - Concept operators: □, ⊔, ¬, ∃, ∀
 - No role operators (only atomic roles)
 - Concept axioms: <u></u>_, ≡
 - No role axioms
- E.g.:

 $ProudParent \equiv Person \sqcap \forall hasChild.(Doctor \sqcup \exists hasChild.Doctor)$









- *S* used for *ALC* extended with (role) transitivity axioms
- Additional letters indicate various extensions, e.g.:
 - \mathcal{H} for role hierarchy (e.g., hasDaughter \sqsubseteq hasChild)
 - \mathcal{R} for role box (e.g., hasParent \circ hasBrother \sqsubseteq hasUncle)
 - O for nominals/singleton classes (e.g., {Italy})
 - \mathcal{I} for inverse roles (e.g., isChildOf \equiv hasChild⁻)
 - \mathcal{N} for number restrictions (e.g., \geq 2hasChild, \leq 3hasChild)
 - Q for qualified number restrictions (e.g., \geq 2hasChild.Doctor)
 - \mathcal{F} for functional number restrictions (e.g., ≤ 1 hasMother)
- E.g., SHIQ = S + role hierarchy + inverse roles + QNRs









- Numerous other extensions have been investigated
 - Concrete domains (numbers, strings, etc)
 - DL-safe rules (Datalog-like rules)
 - Fixpoints
 - Role value maps
 - Additional role constructors (\cap , \cup , \neg , \circ , id, ...)
 - Nary (i.e., predicates with arity >2)
 - Temporal
 - Fuzzy
 - Probabilistic
 - Non-monotonic
 - Higher-order
 - ...

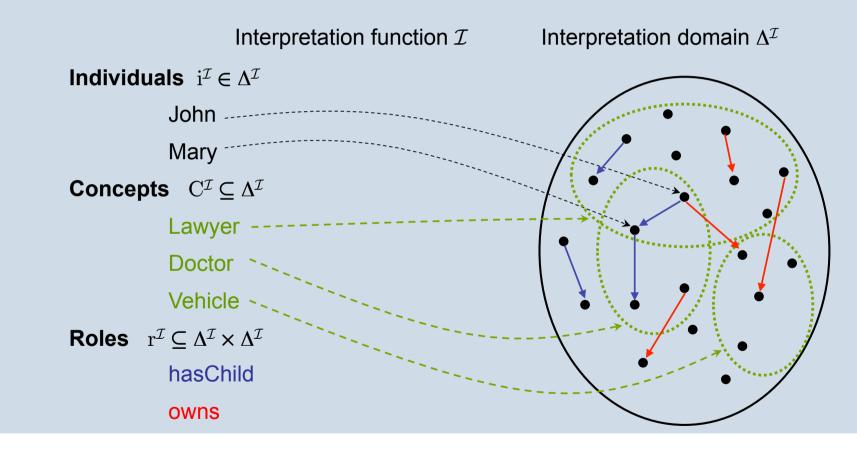








Via translaton to FOL, or directly using FO model theory:











Interpretation function extends to concept expressions in the obvious(ish) way, e.g.:

$$(C \sqcap D)^{\mathcal{I}} = C^{\mathcal{I}} \cap D^{\mathcal{I}}$$
$$(C \sqcup D)^{\mathcal{I}} = C^{\mathcal{I}} \cup D^{\mathcal{I}}$$
$$(\neg C)^{\mathcal{I}} = \Delta^{\mathcal{I}} \setminus C^{\mathcal{I}}$$
$$\{x\}^{\mathcal{I}} = \{x^{\mathcal{I}}\}$$
$$(\exists R.C)^{\mathcal{I}} = \{x \mid \exists y. \langle x, y \rangle \in R^{\mathcal{I}} \land y \in C^{\mathcal{I}}\}$$
$$(\forall R.C)^{\mathcal{I}} = \{x \mid \forall y. (x, y) \in R^{\mathcal{I}} \Rightarrow y \in C^{\mathcal{I}}\}$$
$$(\leqslant nR)^{\mathcal{I}} = \{x \mid \#\{y \mid \langle x, y \rangle \in R^{\mathcal{I}}\} \leqslant n\}$$
$$(\geqslant nR)^{\mathcal{I}} = \{x \mid \#\{y \mid \langle x, y \rangle \in R^{\mathcal{I}}\} \geqslant n\}$$









- Given a model M = $\langle D, \cdot^I \rangle$
 - $M \models C \sqsubseteq D$ iff $C^I \subseteq D^I$
 - $M \models C \equiv D$ iff $C^I = D^I$
 - $M \models C(a)$ iff $a^I \in C^I$
 - $M \models R(a, b)$ iff $\langle a^I, b^I \rangle \in R^I$
 - $M \models \langle \mathcal{T}, \mathcal{A} \rangle$ iff for every axiom $ax \in \mathcal{T} \cup \mathcal{A}, M \models ax$









- Satisfiability and entailment
 - A KB \mathcal{K} is satisfiable iff there exists a model M s.t. M $\models \mathcal{K}$
 - A concept C is satisfiable w.r.t. a KB K iff there exists a model M = (D, ·I) s.t. M ⊨ K and C^I ≠ Ø
 - A KB K entails an axiom ax (written K ⊨ ax) iff for every model M of K, M ⊨ ax (i.e., M ⊨ K implies M ⊨ ax)







- E.g., $\mathcal{T} = \{ \text{Doctor} \sqsubseteq \text{Person}, \text{Parent} \equiv \text{Person} \sqcap \exists \text{hasChild.Person}, \\ \text{HappyParent} \equiv \text{Parent} \sqcap \forall \text{hasChild.(Doctor} \sqcup \exists \text{hasChild.Doctor}) \}$
 - $A = \{$ John:HappyParent, John hasChild Mary, John hasChild Sally, Mary:¬Doctor, Mary hasChild Peter, Mary:(≤ 1 hasChild)
 - $\checkmark \bullet \mathcal{K} \vDash \text{John:Person } ?$
 - $\checkmark \bullet \mathcal{K} \vDash \text{Peter:Doctor } ?$
 - ✓ $\mathcal{K} \models$ Mary:HappyParent ?
 - What if we add "Mary hasChild Jane"?
 - $\mathcal{K} \models \text{Peter} = \text{Jane}$
 - What if we add "HappyPerson \equiv Person $\sqcap \exists$ hasChild.Doctor"?
 - $\mathcal{K} \vDash HappyPerson \sqsubseteq Parent$









DL and FOL

- Most DLs are subsets of C2
 - But reduction to C2 may be (highly) non-trivial
 - Trans(R) naively reduces to $\forall x, y, z. R(x, y) \land R(y, z) \rightarrow R(x, z)$
- Why use DL instead of C2?
 - Syntax is succinct and convenient for KR applications
 - Syntactic conformance guarantees being inside C2
 - Even if reduction to C2 is non-obvious
 - Different combinations of constructors can be selected
 - To guarantee decidability
 - To reduce complexity
 - Decidability/complexity landscape mapped out in great detail
 - See <u>http://www.cs.man.ac.uk/~ezolin/dl/</u>











Complexity of reasoning in Description Logics Note: the information here is (always) incomplete and <u>updated</u> often

Base description logic: Attributive Language with Complements





Concept constructors:	Role constructors:	trans reg
	✓ <i>I</i> - role inverses: R^- ∩ - role intersection ³ : $R \cap S$ ∪ - role union: $R \cup S$ ¬ - role complement: <u>full</u> = o - role chain (composition): RoS = * - reflexive-transitive closure ⁴ : R^* = <i>id</i> - concept identity: <i>id</i> (<i>C</i>) Forbid = complex roles ⁵ in number restrictions ⁶	
 TBox is internalized in extensions of ALCIO, see [76, Lemma 4.12], [54, p.3] Empty TBox Acyclic TBox (A≡C, A is a concept name; no cycles) General TBox (C⊆D for arbitrary concepts C and D) 	Role axioms (RBox): $\ensuremath{\boxtimes}$ S - Role transitivity: Trans(R) $\ensuremath{\boxtimes}$ \mathcal{H} - Role hierarchy: $R \subseteq S$ $\ensuremath{\boxtimes}$ \mathcal{R} - Complex role inclusions: $RoS \subseteq R$, $RoS \subseteq S$ $\ensuremath{\otimes}$ s - some additional features	OWL-Lite OWL-DL OWL 1.1

Reset You have selected the Description Logic: SHOLN						
Complexity of reasoning problems ^Z						
Reasoning problem	Complexity ⁸	Comments and references				
Concept satisfiability	NExpTime-complete	 <u>Hardness</u> of even <i>ALCFIO</i> is proved in [76, Corollary 4.13]. In that paper, the result is formulated for <i>ALCQIO</i>, but only number restrictions of the form (≤1R) are used in the proof. A different proof of the NExpTime-hardness for <i>ALCFIO</i> is given in [54] (even with 1 nominal, and role inverses not used in number restrictions). <u>Upper bound</u> for <i>SHOIQ</i> is proved in [77, Corollary 6.31] with numbers coded in unary (for binary coding, the upper bound remains an open problem for all logics in between <i>ALCNIO</i> and <i>SHOIQ</i>. Important: in number restrictions, only <i>simple</i> roles (i.e. which are neither transitive nor have a transitive subroles) are allowed; otherwise we gain undecidability even in <i>SHN</i>, see [46]. Remark: recently [47] it was observed that, in many cases, one can use transitive roles in number restrictions – and still have a decidable logic! So the above notion of a <i>simple</i> role could be substantially extended. 				
ABox consistency	NExpTime-complete	By reduction to concept satisfiability problem in presence of nominals shown in [69, Theorem 3.7].				









Complexity Measures

- Taxonomic complexity
 - Measured w.r.t. total size of "schema" axioms
- Data complexity
 - Measured w.r.t. total size of "data" facts
- Query complexity
 - Measured w.r.t. size of query
- Combined complexity

Measured w.r.t. total size of KB (plus query if appropriate)









Complexity Classes

- LogSpace, PTime, NP, PSpace, ExpTime, etc
 - worst case for a given problem w.r.t. a given parameter
 - X-hard means at-least this hard (could be harder); in X means no harder than this (could be easier); X-complete means both hard and in, i.e., exactly this hard
 - e.g., SROIQ KB satisfiability is 2NExpTime-complete w.r.t. combined complexity and NP-hard w.r.t. data complexity
- Note that:
 - this is for the worst case, not a typical case
 - complexity of problem means we can never devise a more efficient (in the worst case) algorithm
 - complexity of algorithm may, however, be even higher (in the worst case)









DLs and Ontology Languages

- ₩3C's OWL 2 (like OWL, DAML+OIL & OIL) based on DL
 - OWL 2 based on *SROTQ*, i.e., *ALC* extended with transitive roles, a role box nominals, inverse roles and qualified number restrictions
 - OWL 2 EL based on *EL*
 - OWL 2 QL based on DL-Lite
 - OWL 2 EL based on DLP
 - OWL was based on SHOIN
 - only simple role hierarchy, and unqualified NRs











Class/Concept Constructors

OWL Constructor intersectionOf unionOf complementOf oneOf allValuesFrom someValuesFrom maxCardinality minCardinality

DL Syntax	Example
$C_1 \sqcap \ldots \sqcap C_n$	Human ⊓ Male
$C_1 \sqcup \ldots \sqcup C_n$	Doctor ⊔ Lawyer
$\neg C$	¬Male
$\{x_1\}\sqcup\ldots\sqcup\{x_n\}$	{john} ⊔ {mary}
$\forall P.C$	∀hasChild.Doctor
$\exists P.C$	∃hasChild.Lawyer
$\leqslant nP$	≤1hasChild
$\geqslant nP$	≥2hasChild









Ontology Axioms

OWL Syntax	DL Syntax	Example
subClassOf	$C_1 \sqsubseteq C_2$	Human ⊑ Animal ⊓ Biped
equivalentClass	$C_1 \equiv C_2$	$Man \equiv Human \sqcap Male$
subPropertyOf	$P_1 \sqsubseteq P_2$	hasDaughter ⊑ hasChild
equivalentProperty	$P_1 \equiv P_2$	$cost \equiv price$
transitiveProperty	$P^+ \sqsubseteq P$	ancestor $+ \sqsubseteq$ ancestor

OWL Syntax	DL Syntax	Example
type	a : C	John : Happy-Father
property	$\langle a,b \rangle$: R	$\langle John, Mary \rangle$: has-child

• An Ontology is *usually* considered to be a TBox

but an OWL ontology is a mixed set of TBox and ABox axioms









Other OWL Features

- XSD datatypes and (in OWL 2) facets, e.g.,
 - integer, string and (in OWL 2) real, float, decimal, datetime, ...
 - minExclusive, maxExclusive, length, ...
 - PropertyAssertion(hasAge Meg "17"^^xsd:integer)
 - DatatypeRestriction(xsd:integer xsd:minInclusive "5"^^xsd:integer xsd:maxExclusive "10"^^xsd:integer)

These are equivalent to (a limited form of) **DL concrete domains**

- Keys
 - E.g., HasKey(Vehicle Country LicensePlate)
 - Country + License Plate is a unique identifier for vehicles

This is equivalent to (a limited form of) DL safe rules









Obvious Database Analogy

- Ontology axioms analogous to DB schema
 - Schema describes structure of and constraints on data
- Ontology facts analogous to DB data
 - Instantiates schema
 - Consistent with schema constraints
- But there are also important differences...









Obvious Database Analogy

Database:

- Closed world assumption (CWA)
 - Missing information treated as false
- Unique name assumption (UNA)
 - Each individual has a single, unique name
- Schema behaves as constraints on structure of data
 - Define legal database states

Ontology:

- Open world assumption (OWA)
 - Missing information treated as unknown
- No UNA
 - Individuals may have more than one name
- Ontology axioms behave like implications (inference rules)
 - Entail implicit information











Database -v- Ontology

E.g., given the following **ontology/schema**:

HogwartsStudent \equiv Student $\sqcap \exists$ attendsSchool.HogwartsHogwartsStudent $\sqsubseteq \forall$ hasPet.(Owl or Cat or Toad)hasPet \equiv isPetOf \cdot (i.e., hasPet inverse of isPetOf) \exists hasPet. $\top \sqsubseteq$ Human(i.e., domain of hasPet is Human)Phoenix $\sqsubseteq \forall$ isPetOf.Wizard(i.e., only Wizards have Phoenix pets)

(i.e., Muggles and Wizards are disjoint)



Muggle $\sqsubseteq \neg$ Wizard





Database -v- Ontology

And the following facts/data:

HarryPotter: Wizard DracoMalfoy: Wizard HarryPotter hasFriend RonWeasley HarryPotter hasFriend HermioneGranger HarryPotter hasPet Hedwig

Query: Is Draco Malfoy a friend of HarryPotter?

- DB: No
- Ontology: Don't Know

OWA (didn't say Draco was not Harry's friend)









And the following facts/data:

HarryPotter: Wizard DracoMalfoy: Wizard HarryPotter hasFriend RonWeasley HarryPotter hasFriend HermioneGranger HarryPotter hasPet Hedwig

Query: How many friends does Harry Potter have?

- DB: 2
- Ontology: at least 1

No UNA (Ron and Hermione may be 2 names for same person)









And the following facts/data:

HarryPotter: Wizard DracoMalfoy: Wizard HarryPotter hasFriend RonWeasley HarryPotter hasFriend HermioneGranger HarryPotter hasPet Hedwig

→ RonWeasley ≠ HermioneGranger

Query: How many friends does Harry Potter have?

- DB: 2
- Ontology: at least 2

OWA (Harry may have more friends we didn't mention yet)









And the following facts/data:

HarryPotter: Wizard DracoMalfoy: Wizard HarryPotter hasFriend RonWeasley HarryPotter hasFriend HermioneGranger HarryPotter hasPet Hedwig RonWeasley ≠ HermioneGranger

→ HarryPotter: ∀hasFriend.{RonWeasley} ⊔ {HermioneGranger}

- Query: How many friends does Harry Potter have?
 - DB: 2
 - Ontology: 2!









Inserting new facts/data:

Dumbledore: WizardFawkes: PhoenixFawkes isPetOf Dumbledore

What is the response from DBMS?

Update rejected: constraint violation

Domain of hasPet is Human; Dumbledore is not Human (CWA)

What is the response from Ontology reasoner?

- Infer that Dumbledore is Human (domain restriction)
- Also infer that Dumbledore is a Wizard (only a Wizard can have a pheonix as a pet)









 \exists hasPet. $\top \sqsubseteq$ Human Phoenix $\sqsubseteq \forall$ isPetOf.Wizard

DB Query Answering

- Schema plays no role
 - Data must explicitly satisfy schema constraints
- Query answering amounts to model checking
 - I.e., a "look-up" against the data
- Can be very efficiently implemented
 - Worst case complexity is low (logspace) w.r.t. size of data









Ontology Query Answering

- Ontology axioms play a powerful and crucial role
 - Answer may include implicitly derived facts
 - Can answer conceptual as well as extensional queries
 - E.g., Can a Muggle have a Phoenix for a pet?
- Query answering amounts to theorem proving
 - I.e., logical entailment
- May have very high worst case complexity
 - E.g., for OWL, NP-hard w.r.t. size of data (upper bound is an open problem)
 - Implementations may still behave well in typical cases
 - Fragments/profiles may have much better complexity









Ontology Based Information Systems

- Analogous to relational database management systems
 - Ontology \approx schema; instances \approx data
- Some important (dis)advantages
 - + (Relatively) easy to maintain and update schema
 - Schema plus data are integrated in a logical theory
 - + Query answers reflect both schema and data
 - + Can deal with incomplete information
 - + Able to answer both intensional and extensional queries
 - Semantics can seem counter-intuitive, particularly w.r.t. data
 - Open -v- closed world; axioms -v- constraints
 - Query answering (logical entailment) may be much more difficult
 - Can lead to scalability problems with expressive logics









Ontology Based Information Systems











Back to our Scheduled Program





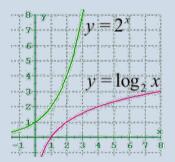




Theory ~> Practice

- 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
- OWL is decidable, but highly highly intractable
 - N2ExpTime-comlete combined complexity
 - NP-hard data complexity (-v- logspace for databases)





How can we provide robustly scalable query answering?









Various Approaches — Different Tradeoffs

1 Use full power of OWL and a complete reasoner:

- \checkmark 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









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff

 $KB \cup \{x: (HeartDisease \sqcap \neg VascularDisease)\}$ is *not* satisfiable









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model

 $x: HeartDisease \sqcap \neg VascularDisease$







Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model

x: HeartDisease $\square \neg$ VascularDisease x: HeartDisease









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model

x: HeartDisease □ ¬VascularDisease *x*: HeartDisease









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model
- $x: HeartDisease \sqcap \neg VascularDisease$
- x: HeartDisease
- x: Disease
- $x: \exists affects. Heart$









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model
- x: HeartDisease $\sqcap \neg$ VascularDisease
- x: HeartDisease
- \boldsymbol{x} : Disease
- $x: \exists affects. Heart$









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model
- x: HeartDisease □ ¬VascularDisease x: HeartDisease x: Disease x: ∃affects.Heart (x, y): affects y: Heart









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model
- x: HeartDisease □ ¬VascularDisease x: HeartDisease x: Disease x: ∃affects.Heart (x, y): affects y: Heart









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model
- x: HeartDisease □ ¬VascularDisease x: HeartDisease x: Disease x: ∃affects.Heart (x, y): affects y: Heart y: MuscularOrgan y: ∃isPartOf.CirculatorySystem









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model
- #: HeartDisease □ ¬VascularDisease #: HeartDisease #: Disease #: ∃affects.Heart (x, y) : affects y : Heart y : MuscularOrgan y : ∃isPartOf.CirculatorySystem









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model

```
#: HeartDisease T -VascularDisease
#: HeartDisease
#: Disease
#: Jaffects.Heart
(x, y) : affects
y : Heart
y : MuscularOrgan
y : JisPartOf.CirculatorySystem
(y, z) : isPartOf
z : CirculatorySystem
```









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 KB ∪ {x:(HeartDisease □ ¬VascularDisease)} is not satisfiable
- Algorithm tries to construct (an abstraction of) a model
- $x: \mathsf{HeartDisease}$
- $x: \mathsf{Disease}$
- $x: \exists affects. Heart$
- (x, y) : affects
 - y: Heart
 - y: MuscularOrgan
 - y: ∃isPartOf.CirculatorySystem
- (y, z): isPartOf
 - *z* : CirculatorySystem









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff KB \cup {x:(HeartDisease $\sqcap \neg$ VascularDisease)} is *not* satisfiable
- Algorithm tries to construct (an abstraction of) a model

x: Heart Disease $\square \neg Vascular Disease \qquad x: \neg Vascular Disease$

- m: Heart Disease
- x: Disease
- m: Haffects. Heart
- (x, y) : affects
 - u: Heart
 - y: MuscularOrgan
 - y: ∃isPartOf.CirculatorySystem
- (y, z): isPartOf
 - z: CirculatorySystem









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff KB \cup {x:(HeartDisease $\sqcap \neg$ VascularDisease)} is *not* satisfiable
- Algorithm tries to construct (an abstraction of) a model

x: Heart Disease $\Box \neg Vascular Disease$ $x: \neg Vascular Disease$

- m: Heart Disease
- x: Disease
- m: Haffects. Heart
- (x, y) : affects
 - u: Heart
 - y: MuscularOrgan
 - y: ∃isPartOf.CirculatorySystem
- (y, z): isPartOf
 - z: CirculatorySystem









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - KB \cup {x:(HeartDisease $\sqcap \neg$ VascularDisease)} is *not* satisfiable
- Algorithm tries to construct (an abstraction of) a model
- x: Heart Disease $\square \neg Vascular Disease x: \neg Vascular Disease$
- m: Heart Disease
- x: Disease
- m: Haffects. Heart

(x, y) : affects

- u: Heart
- y: MuscularOrgan
- y: ∃isPartOf.CirculatorySystem
- (y, z): isPartOf
 - z: CirculatorySystem









- $x: \neg \mathsf{Disease} \sqcup$
 - -Jaffects.(JisPartOf.CirculatorySystem)

Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - KB \cup {x:(HeartDisease $\sqcap \neg$ VascularDisease)} is *not* satisfiable
- Algorithm tries to construct (an abstraction of) a model

x: Heart Disease $\square \neg$ Vascular Disease $x: \neg$ Vascular Disease m: Heart Disease

x: Disease

m: Haffects. Heart

(x, y) : affects

u: Heart

- y: MuscularOrgan
- y: ∃isPartOf.CirculatorySystem

(y, z): isPartOf

z: CirculatorySystem



Information Systems Group







 $x: \neg \mathsf{Disease} \sqcup$

-Jaffects.(JisPartOf.CirculatorySystem)

Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $\mathsf{KB} \cup \{ x: (\mathsf{HeartDisease} \sqcap \neg \mathsf{VascularDisease}) \} \text{ is } \textit{not} \text{ satisfiable}$
- Algorithm tries to construct (an abstraction of) a model

x: HeartDisease □ ¬VascularDisease x: HeartDisease x: Disease x: ∃affects.Heart (x, y): affects y: Heart y: MuscularOrgan y: ∃isPartOf.CirculatorySystem (y, z): isPartOf

z : CirculatorySystem



Information Systems Group







Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $KB \cup \{x: (HeartDisease \sqcap \neg VascularDisease)\}$ is *not* satisfiable
- Algorithm tries to construct (an abstraction of) a model

x: HeartDisease $\square \neg VascularDisease$ x: HeartDisease x: Disease

```
#: Baffects.Heart
```

```
(x,y) : affects
```

y : Heart

- y: MuscularOrgan
- y: ∃isPartOf.CirculatorySystem

(y, z): isPartOf

z : CirculatorySystem



Information Systems Group







#:¬VascularDisease #:¬Disease ⊔ ¬∃affects.(∃isPartOf.CirculatorySystem) #:¬Disease

Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - KB \cup {x:(HeartDisease $\sqcap \neg$ VascularDisease)} is *not* satisfiable
- Algorithm tries to construct (an abstraction of) a model

x: Heart Disease $\Box \neg Vascular Disease \qquad x: \neg Vascular Disease$ m: Heart Disease x: Diseasem: Haffects. Heart (x, y) : affects

- u: Heart
- y: MuscularOrgan
- y: ∃isPartOf.CirculatorySystem

(y, z): isPartOf

z: CirculatorySystem



Information Systems Group







 $x: \neg \mathsf{Disease} \sqcup$

Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $\mathsf{KB} \cup \{ x: (\mathsf{HeartDisease} \sqcap \neg \mathsf{VascularDisease}) \} \text{ is } \textit{not} \text{ satisfiable}$

x : ¬VascularDisease

Algorithm tries to construct (an abstraction of) a model

x: HeartDisease □ ¬VascularDisease x: HeartDisease x: Disease x: ∃affects.Heart (x, y): affects y: Heart y: MuscularOrgan

- y: ∃isPartOf.CirculatorySystem
- (y, z): isPartOf
 - *z* : CirculatorySystem



Information Systems Group







#:¬Disease⊔
 ¬∃affects.(∃isPartOf.CirculatorySystem)
#:¬∃affects.(∃isPartOf.CirculatorySystem)

Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $\mathsf{KB} \cup \{ x: (\mathsf{HeartDisease} \sqcap \neg \mathsf{VascularDisease}) \} \text{ is } \textit{not} \text{ satisfiable}$
- Algorithm tries to construct (an abstraction of) a model
- x: HeartDisease □ ¬VascularDisease x: HeartDisease x: Disease x: ∃affects.Heart (x, y): affects
 - y:Heart
 - y: MuscularOrgan
 - y: ∃isPartOf.CirculatorySystem
- (y, z): isPartOf
 - *z* : CirculatorySystem



Information Systems Group







#:¬VascularDisease #:¬Disease□ ¬∃affects.(∃isPartOf.CirculatorySystem) #:¬∃affects.(∃isPartOf.CirculatorySystem)

Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $\mathsf{KB} \cup \{x: (\mathsf{HeartDisease} \sqcap \neg \mathsf{VascularDisease})\} \text{ is } \textit{not} \text{ satisfiable}$
- Algorithm tries to construct (an abstraction of) a model

 $x: \exists affects. Heart$

(x, y) : affects

y : Heart

y: MuscularOrgan

 $y: \exists isPartOf.CirculatorySystem$

(y, z): isPartOf

z : CirculatorySystem



Information Systems Group







#:¬VascularDisease
#:¬Disease□
¬∃affects.(∃isPartOf.CirculatorySystem)
#:¬∃affects.(∃isPartOf.CirculatorySystem)
#:∀affects.(∀isPartOf.¬CirculatorySystem)

Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $\mathsf{KB} \cup \{ x: (\mathsf{HeartDisease} \sqcap \neg \mathsf{VascularDisease}) \} \text{ is } \textit{not} \text{ satisfiable}$
- Algorithm tries to construct (an abstraction of) a model
- x: HeartDisease $\square \neg$ VascularDisease
- x: Heart Disease
- $\boldsymbol{x}:\mathsf{Disease}$
- $x: \exists affects. Heart$

(x, y) : affects

- y : Heart
- y: MuscularOrgan
- $y: \exists isPartOf.CirculatorySystem$
- (y, z): isPartOf
 - *z* : CirculatorySystem



Information Systems Group







#:¬VascularDisease #:¬Disease⊔ ¬∃affects.(∃isPartOf.CirculatorySystem) #:¬∃affects.(∃isPartOf.CirculatorySystem) #:∀affects.(∀isPartOf.¬CirculatorySystem)

Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $\mathsf{KB} \cup \{x: (\mathsf{HeartDisease} \sqcap \neg \mathsf{VascularDisease})\} \text{ is } \textit{not} \text{ satisfiable}$
- Algorithm tries to construct (an abstraction of) a model
- x: HeartDisease $\sqcap \neg$ VascularDisease
- x: Heart Disease
- $x: \mathsf{Disease}$
- x: ∃affects.Heart

(x, y) : affects

- y : Heart
- y: MuscularOrgan
- $y: \exists isPartOf.CirculatorySystem$
- (y, z): isPartOf
 - z : CirculatorySystem









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $\mathsf{KB} \cup \{x: (\mathsf{HeartDisease} \sqcap \neg \mathsf{VascularDisease})\} \text{ is } \textit{not} \text{ satisfiable}$
- Algorithm tries to construct (an abstraction of) a model

x: HeartDisease $\square \neg$ VascularDisease x: HeartDisease

x : Disease

 $x: \exists affects. Heart$

(x, y) : affects

y : Heart

- y: MuscularOrgan
- $y: \exists isPartOf.CirculatorySystem$
- (y, z): isPartOf

z : CirculatorySystem

#:¬VascularDisease
#:¬Disease□
¬∃affects.(∃isPartOf.CirculatorySystem)
#:¬∃affects.(∃isPartOf.CirculatorySystem)
#:∀affects.(∀isPartOf.¬CirculatorySystem)
#:∀isPartOf.¬CirculatorySystem









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $\mathsf{KB} \cup \{x: (\mathsf{HeartDisease} \sqcap \neg \mathsf{VascularDisease})\} \text{ is } \textit{not} \text{ satisfiable}$
- Algorithm tries to construct (an abstraction of) a model

x: HeartDisease $\square \neg$ VascularDisease x: HeartDisease

x : Disease

 $x: \exists affects. Heart$

(x, y) : affects

y:Heart

- y: MuscularOrgan
- $y: \exists isPartOf.CirculatorySystem$

(y, z): isPartOf

z : CirculatorySystem

#:¬VascularDisease #:¬Disease⊔ ¬∃affects.(∃isPartOf.CirculatorySystem) #:¬∃affects.(∃isPartOf.CirculatorySystem) #:∀affects.(∀isPartOf.¬CirculatorySystem) #:∀isPartOf.¬CirculatorySystem #:¬CirculatorySystem









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $\mathsf{KB} \cup \{ x: (\mathsf{HeartDisease} \sqcap \neg \mathsf{VascularDisease}) \} \text{ is } \textit{not} \text{ satisfiable}$
- Algorithm tries to construct (an abstraction of) a model

x: HeartDisease □ ¬VascularDisease *x*: HeartDisease *x*: Disease *x*: ∃affects.Heart

(x, y) : affects

y : Heart

- y: MuscularOrgan
- $y: \exists isPartOf.CirculatorySystem$

(y, z): isPartOf

z : CirculatorySystem

#:¬VascularDisease #:¬Disease□

- ¬∃affects.(∃isPartOf.CirculatorySystem)
- #: ¬∃affects.(∃isPartOf.CirculatorySystem)
- #: Vaffects.(VisPartOf.¬CirculatorySystem)
- $y: \forall isPartOf. \neg CirculatorySystem$
- ℤ: ¬CirculatorySystem









Standard technique based on (hyper-) tableau

- Reasoning tasks reducible to (un)satisfiability
 - E.g., KB ⊨ HeartDisease ⊑ VascularDisease iff
 - $KB \cup \{x: (HeartDisease \sqcap \neg VascularDisease)\}$ is *not* satisfiable

Algorithm tries to construct (an abstraction of) a model

x: HeartDisease $\square \neg VascularDisease$ x: HeartDisease

x : ¬VascularDisease *x* : ¬Disease ⊔

ℤ: ¬CirculatorySystem

x: Dipage *x*: Banote similarity affects. (BisPartOf. CirculatorySystem) *x*: Banote similarity affects. (BisPartOf. CirculatorySystem) *x*: Banote similarity affects. (BisPartOf. CirculatorySystem) *x*: Banote similarity affects. (BisPartOf. CirculatorySystem)

(x, y) : affects

y:Heart

y: MuscularOrgan

y: ∃isPartOf.CirculatorySystem

(y, z): isPartOf

z : CirculatorySystem



Information Systems Group





y: VisPartOf.¬CirculatorySystem



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 *EL*⁺⁺
 - PTime-complete for combined and data complexity
- OWL 2 QL
 - Based on DL-Lite
 - AC⁰ 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



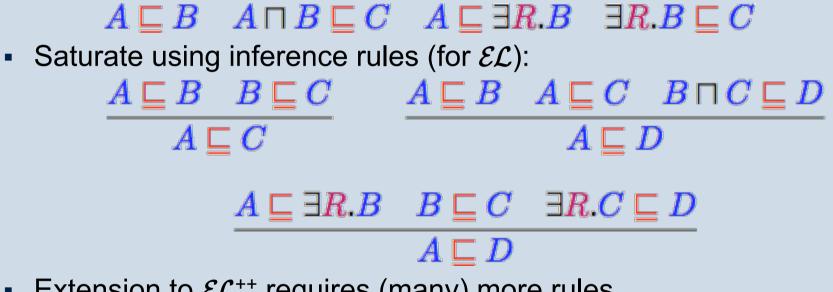






Consequence Based — How Does It Work?

Normalise ontology axioms to standard form:



Extension to *EL*⁺⁺ requires (many) more rules









 $OrganTransplant \equiv Transplant \sqcap \exists site.Organ$ HeartTransplant $\equiv Transplant \sqcap \exists site.Heart$ Heart $\sqsubseteq Organ$









 $OrganTransplant \equiv Transplant \sqcap \exists site.Organ$

 $HeartTransplant \equiv Transplant \sqcap \exists site. Heart \\ Heart \sqsubseteq Organ$









 $OrganTransplant \equiv Transplant \sqcap \exists site. Organ$

 $\begin{aligned} \text{HeartTransplant} \equiv \text{Transplant} \sqcap \exists \textbf{site}. \text{Heart} \\ \text{Heart} \sqsubseteq \text{Organ} \end{aligned}$

 $OrganTransplant \sqsubseteq Transplant$ $OrganTransplant \sqsubseteq \exists site.Organ$









 $OrganTransplant \equiv Transplant \sqcap \exists site.Organ$

 $\begin{aligned} \mathsf{HeartTransplant} &\equiv \mathsf{Transplant} \sqcap \exists \mathsf{site}.\mathsf{Heart} \\ \mathsf{Heart} \sqsubseteq \mathsf{Organ} \end{aligned}$

OrganTransplant ⊑ Transplant OrganTransplant ⊑ ∃site.Organ ∃site.Organ ⊑ SO









 $OrganTransplant \equiv Transplant \sqcap \exists site.Organ$

 $\begin{aligned} \mathsf{HeartTransplant} &\equiv \mathsf{Transplant} \sqcap \exists \mathsf{site}.\mathsf{Heart} \\ \mathsf{Heart} \sqsubseteq \mathsf{Organ} \end{aligned}$

OrganTransplant ⊑ Transplant OrganTransplant ⊑ ∃site.Organ ∃site.Organ ⊑ SO Transplant □ SO ⊑ OrganTransplant









 $OrganTransplant \equiv Transplant \sqcap \exists site.Organ$ HeartTransplant $\equiv Transplant \sqcap \exists site.Heart$ Heart $\sqsubseteq Organ$

OrganTransplant \sqsubseteq Transplant OrganTransplant \sqsubseteq **3site**.Organ **3site**.Organ \sqsubseteq SO Transplant \sqcap SO \sqsubseteq OrganTransplant









 $OrganTransplant \equiv Transplant \sqcap \exists site.Organ$ HeartTransplant \equiv Transplant $\sqcap \exists site.Heart$ Heart $\sqsubseteq Organ$

OrganTransplant \sqsubseteq Transplant OrganTransplant \sqsubseteq \exists site.Organ \exists site.Organ \sqsubseteq SO Transplant \sqcap SO \sqsubseteq OrganTransplant HeartTransplant \sqsubseteq Transplant HeartTransplant \sqsubseteq \exists site.Heart









```
OrganTransplant \equiv Transplant \sqcap \exists site.Organ
HeartTransplant \equiv Transplant \sqcap \exists site.Heart
Heart \sqsubseteq Organ
```

OrganTransplant \sqsubseteq Transplant OrganTransplant \sqsubseteq 3site.Organ \exists site.Organ \sqsubseteq SO Transplant \sqcap SO \sqsubseteq OrganTransplant HeartTransplant \sqsubseteq Transplant HeartTransplant \sqsubseteq 3site.Heart \exists site.Heart \sqsubseteq SH Transplant \sqcap SH \sqsubseteq HeartTransplant









 $OrganTransplant \equiv Transplant \sqcap \exists site.Organ$ HeartTransplant $\equiv Transplant \sqcap \exists site.Heart$ Heart $\sqsubseteq Organ$

OrganTransplant \sqsubseteq Transplant OrganTransplant \sqsubseteq 3site.Organ Bsite.Organ \sqsubseteq SO Transplant \sqcap SO \sqsubseteq OrganTransplant HeartTransplant \sqsubseteq Transplant HeartTransplant \sqsubseteq 3site.Heart Bsite.Heart \sqsubseteq SH Transplant \sqcap SH \sqsubseteq HeartTransplant









 $OrganTransplant \equiv Transplant \sqcap \exists site.Organ$ HeartTransplant $\equiv Transplant \sqcap \exists site.Heart$ Heart $\sqsubseteq Organ$

OrganTransplant \sqsubseteq Transplant OrganTransplant \sqsubseteq 3site.Organ \exists site.Organ \sqsubseteq SO Transplant \sqcap SO \sqsubseteq OrganTransplant HeartTransplant \sqsubseteq Transplant HeartTransplant \sqsubseteq 3site.Heart \exists site.Heart \sqsubseteq SH Transplant \sqcap SH \sqsubseteq HeartTransplant Heart \sqsubseteq Organ







 $OrganTransplant \equiv Transplant \sqcap \exists site.Organ$ HeartTransplant \equiv Transplant $\sqcap \exists site.Heart$ Heart $\sqsubseteq Organ$

OrganTransplant \sqsubseteq Transplant OrganTransplant \sqsubseteq 3site.Organ \exists site.Organ \sqsubseteq SO Transplant \sqcap SO \sqsubseteq OrganTransplant HeartTransplant \sqsubseteq Transplant HeartTransplant \sqsubseteq 3site.Heart \exists site.Heart \sqsubseteq SH Transplant \sqcap SH \sqsubseteq HeartTransplant Heart \sqsubseteq Organ







 $OrganTransplant \equiv Transplant \sqcap \exists site.Organ$ HeartTransplant $\equiv Transplant \sqcap \exists site.Heart$ Heart $\sqsubseteq Organ$

OrganTransplant \sqsubseteq Transplant OrganTransplant \sqsubseteq 3site.Organ Bite.Organ \sqsubseteq SO Transplant \sqcap SO \sqsubseteq OrganTransplant HeartTransplant \sqsubseteq Transplant HeartTransplant \sqsubseteq 3site.Heart Bite.Heart \sqsubseteq SH Transplant \sqcap SH \sqsubseteq HeartTransplant Heart \sqsubseteq Organ



Information Systems Group





 $\frac{A \sqsubseteq \exists R.B \quad B \sqsubseteq C \quad \exists R.C \sqsubseteq D}{A \sqsubseteq D}$

OrganTransplant ≡ Transplant $\sqcap \exists$ site.Organ HeartTransplant ≡ Transplant $\sqcap \exists$ site.Heart Heart ⊑ Organ

OrganTransplant \sqsubseteq Transplant OrganTransplant \sqsubseteq 3site.Organ \exists site.Organ \sqsubseteq SO Transplant \sqcap SO \sqsubseteq OrganTransplant HeartTransplant \sqsubseteq Transplant HeartTransplant \sqsubseteq 3site.Heart \exists site.Heart \sqsubseteq SH Transplant \sqcap SH \sqsubseteq HeartTransplant Heart \sqsubseteq Organ $\frac{A \sqsubseteq \exists R.B \quad B \sqsubseteq C \quad \exists R.C \sqsubseteq D}{A \sqsubseteq D}$

 $HeartTransplant \sqsubseteq SO$









OrganTransplant ≡ Transplant $\sqcap \exists$ site.Organ HeartTransplant ≡ Transplant $\sqcap \exists$ site.Heart Heart ⊑ Organ

OrganTransplant \sqsubseteq Transplant OrganTransplant \sqsubseteq 3site.Organ Bsite.Organ \sqsubseteq SO Transplant \sqcap SO \sqsubseteq OrganTransplant HeartTransplant \sqsubseteq Transplant HeartTransplant \sqsubseteq 3site.Heart Bsite.Heart \sqsubseteq SH Transplant \sqcap SH \sqsubseteq HeartTransplant Heart \sqsubseteq Organ $\frac{A \sqsubseteq B \quad A \sqsubseteq C \quad B \sqcap C \sqsubseteq D}{A \sqsubseteq D}$

 $HeartTransplant \sqsubseteq SO$









OrganTransplant ≡ Transplant $\sqcap \exists$ site.Organ HeartTransplant ≡ Transplant $\sqcap \exists$ site.Heart Heart ⊑ Organ

OrganTransplant \sqsubseteq Transplant OrganTransplant \sqsubseteq 3site.Organ Bite.Organ \sqsubseteq SO Transplant \sqcap SO \sqsubseteq OrganTransplant HeartTransplant \sqsubseteq Transplant HeartTransplant \sqsubseteq 3site.Heart Bite.Heart \sqsubseteq SH Transplant \sqcap SH \sqsubseteq HeartTransplant Heart \sqsubseteq Organ $\frac{A \sqsubseteq B \quad A \sqsubseteq C \quad B \sqcap C \sqsubseteq D}{A \sqsubseteq D}$

 $\begin{array}{l} \mathsf{HeartTransplant}\sqsubseteq\mathsf{SO}\\ \mathsf{HeartTransplant}\sqsubseteq\mathsf{OrganTransplant}\\ \end{array}$









Schema Reasoning — Solved Problem?

S	NOMED CT	GALEN	FMA	GO	
Logic	EL	EL	EL	EL	
#classes	$315,\!489$	$23,\!136$	78,977	19,468	
#properties	58	950	7	1	
#axioms	$430,\!844$	$36,\!547$	121,712	$28,\!897$	
#⊑	$> 10^{11}$	$> 10^{8}$	$> 10^{9}$	$> 10^{8}$	
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	SHIF	SHOIN	ALCH	SHOIN	
#classes	$19,\!145$	1,728	70,576	118	
#properties	82	145	189	264	
#axioms	35,770	2,419	100,304	265	
#⊑	$> 10^{8}$	$> 10^{6}$	$> 10^{9}$	$> 10^4$	
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{\varSigma^{\mathcal{L}}}]} $	Classif. time (seconds)			
				HermiT	MORe		
					total	HermiT	ELK
GO	0	100%	0%	7.1	$2.2~(\downarrow 69.0\%)$	0	0.1
Gazeteer	0	100%	0%	838.1	$28.2 (\downarrow 96.6\%)$	0	15.6
NCI	65	94.9%	15.4%	84.1	$28.6 (\downarrow 66.0\%)$	15.8	3.3
Protein	12	98.1%	6.6%	11.4	$2.9~(\downarrow 74.6\%)$	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	$37.6 (\downarrow 65.8\%)$	17.6	3.2
NCI+Mouse	65	96.0%	13.3%	93.7	$31.0 (\downarrow 66.9\%)$	16.6	2.6









OWL 2 EL — Data Retrieval Queries?

PTime potentially problematical for very large datasets









OWL 2 EL — Data Retrieval Queries?

- 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, Zakharyaschev: 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

- **2** Use a suitable "profile" and specialised reasoner:
- ✓ LogSpace query answering (in size of data)
- ✓ Reliable answers (for inputs in the profile)
- Restricted expressivity of the ontology language
- × Reasoners reject inputs outside profile









Various Approaches — Different Tradeoffs

2 Use a suitable "profile" and specialised reasoner:

- ✓ LogSpace query answering (in size of data)
- ✓ Reliable answers (for inputs in the profile)
- Restricted expressivity of the ontology language
- × Reasoners reject inputs outside profile

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)









Given ontology \mathcal{O} query \mathcal{Q} and mappings \mathcal{M} :



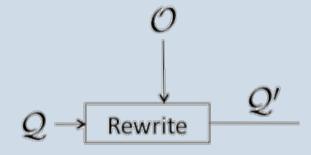






Given ontology \mathcal{O} query \mathcal{Q} and mappings \mathcal{M} :

 Rewrite Q → Q' s.t. answering Q' without O equivalent to answering Q w.r.t. O for any dataset



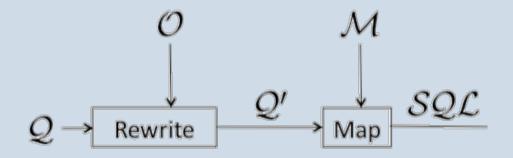






Given ontology \mathcal{O} query \mathcal{Q} and mappings \mathcal{M} :

- Rewrite Q → Q' s.t. answering Q' without O equivalent to answering Q w.r.t. O for any dataset
- Map ontology queries → DB queries (typically SQL) using mappings *M* to rewrite *Q*' into a DB query



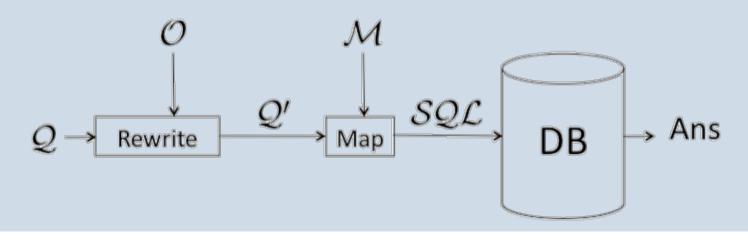






Given ontology \mathcal{O} query \mathcal{Q} and mappings \mathcal{M} :

- Rewrite Q → Q' s.t. answering Q' without O equivalent to answering Q w.r.t. O for any dataset
- Map ontology queries → DB queries (typically SQL) using mappings *M* to rewrite *Q*' into a DB query
- Evaluate (SQL) query against DB











Query Rewriting — Example

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

 $Q \quad Q(x) \leftarrow \mathsf{treats}(x, y) \land \mathsf{Patient}(y)$

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









Query Rewriting — Example

- $\mathcal{O} \left\{ \begin{array}{c} \mathsf{Doctor} \sqsubseteq \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consultant} \sqsubseteq \mathsf{Doctor} \end{array} \right.$
- $\mathcal{Q} \quad Q(x) \leftarrow \mathsf{treats}(x, y) \land \mathsf{Patient}(y)$

 $\mathcal{Q}' \begin{cases} Q(x) \leftarrow \mathsf{treats}(x, y) \land \mathsf{Patient}(y) \\ Q(x) \leftarrow \mathsf{Doctor}(x) \land \mathsf{Patient}(f(x)) \\ Q(x) \leftarrow \mathsf{treats}(x, f(x)) \land \mathsf{Doctor}(x) \\ Q(x) \leftarrow \mathsf{Doctor}(x) \\ Q(x) \leftarrow \mathsf{Consultant}(x) \end{cases}$

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







Query Rewriting — Example

- $\mathcal{O} \left\{ \begin{array}{c} \mathsf{Doctor} \sqsubseteq \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consultant} \sqsubseteq \mathsf{Doctor} \end{array} \right.$
- Q $Q(x) \leftarrow \text{treats}(x, y) \land \text{Patient}(y)$

$$\mathcal{Q}' \begin{cases} Q(x) \leftarrow \mathsf{treats}(x, y) \land \mathsf{Patient}(y) \\ \hline Q(x) \leftarrow \mathsf{Doctor}(x) \land \mathsf{Patient}(f(x)) \\ \hline Q(x) \leftarrow \mathsf{treats}(x, f(x)) \land \mathsf{Doctor}(x) \\ Q(x) \leftarrow \mathsf{Doctor}(x) \\ Q(x) \leftarrow \mathsf{Consultant}(x) \end{cases}$$

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









Query Rewriting — Example

- $\mathcal{O} \left\{ \begin{array}{c} \mathsf{Doctor} \sqsubseteq \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consultant} \sqsubseteq \mathsf{Doctor} \end{array} \right.$
- Q $Q(x) \leftarrow \text{treats}(x, y) \land \text{Patient}(y)$

$$\mathcal{Q}' \begin{cases} Q(x) \leftarrow \mathsf{treats}(x, y) \land \mathsf{Patient}(y) \\ Q(x) \leftarrow \mathsf{Doctor}(x) \land \mathsf{Patient}(f(x)) \\ Q(x) \leftarrow \mathsf{treats}(x, f(x)) \land \mathsf{Doctor}(x) \\ Q(x) \leftarrow \mathsf{Doctor}(x) \\ Q(x) \leftarrow \mathsf{Doctor}(x) \\ Q(x) \leftarrow \mathsf{Consultant}(x) \end{cases}$$

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









Query Rewriting — Example

- $\mathcal{O} \left\{ \begin{array}{c} \mathsf{Doctor} \sqsubseteq \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consultant} \sqsubseteq \mathsf{Doctor} \end{array} \right.$
- $\mathcal{Q} \quad Q(x) \leftarrow \mathsf{treats}(x, y) \land \mathsf{Patient}(y)$

$$\mathcal{Q}' \begin{cases} Q(x) \leftarrow \mathsf{treats}(x, y) \land \mathsf{Patient}(y) \\ \hline Q(x) \leftarrow \mathsf{Doctor}(x) \land \mathsf{Patient}(f(x)) \\ \hline Q(x) \leftarrow \mathsf{treats}(x, f(x)) \land \mathsf{Doctor}(x) \\ Q(x) \leftarrow \mathsf{Doctor}(x) \\ \hline Q(x) \leftarrow \mathsf{Consultant}(x) \end{cases}$$

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

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







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:

- \checkmark 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









Materialisation — How Does It Work?

Given (RDF) data DB, ontology \mathcal{O} and query \mathcal{Q} :







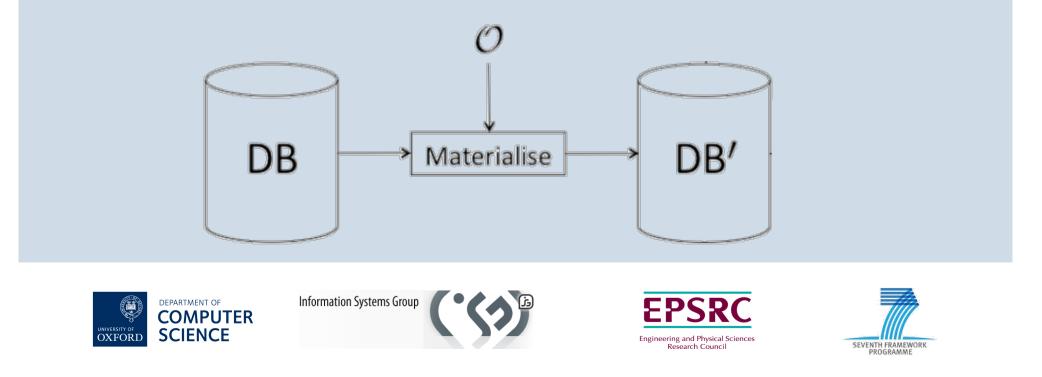


Materialisation — How Does It Work?

Given (RDF) data DB, ontology \mathcal{O} and query \mathcal{Q} :

• Materialise (RDF) data DB \rightarrow DB' s.t. evaluating Q w.r.t. DB' equivalent to answering Q w.r.t. DB and O

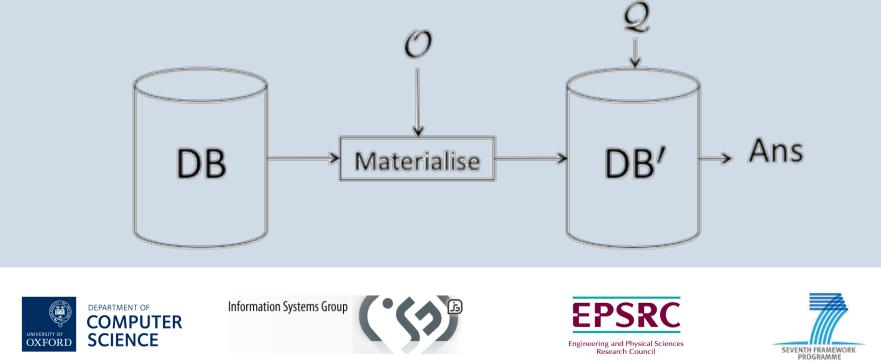
nb: Closely related to chase procedure used with DB dependencies



Materialisation — How Does It Work?

Given (RDF) data DB, ontology \mathcal{O} and query \mathcal{Q} :

- Materialise (RDF) data DB → DB' s.t. evaluating Q w.r.t. DB' equivalent to answering Q w.r.t. DB and O
 nb: Closely related to chase procedure used with DB dependencies
- Evaluate Q against DB'



```
\mathcal{O} \left\{ \begin{array}{l} \mathsf{Doctor} \equiv \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consulatant} \sqsubseteq \mathsf{Doctor} \end{array} \right. \\ \mathsf{DB} \left\{ \begin{array}{l} \mathsf{treats}(d_1, p_1) \\ \mathsf{Patient}(p_1) \\ \mathsf{Doctor}(d_2) \\ \mathsf{Consultant}(c_1) \end{array} \right. \\ \end{array} \right.
```









 $\mathcal{O} \left\{ \begin{array}{l} \mathsf{Doctor} \equiv \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consulatant} \sqsubseteq \mathsf{Doctor} \end{array} \right. \\ \mathsf{DB} \left\{ \begin{array}{l} \mathsf{treats}(d_1, p_1) \\ \mathsf{Patient}(p_1) \\ \mathsf{Doctor}(d_2) \\ \mathsf{Consultant}(c_1) \end{array} \right. \\ \mathsf{DB}' \left\{ \begin{array}{l} \mathsf{treats}(d_1, p_1) \\ \mathsf{Patient}(p_1) \\ \mathsf{Doctor}(d_2) \\ \mathsf{Consultant}(c_1) \\ \mathsf{Doctor}(d_1) \\ \mathsf{Doctor}(c_1) \end{array} \right. \\ \mathsf{DB}' \left\{ \begin{array}{l} \mathsf{treats}(d_1, p_1) \\ \mathsf{Patient}(p_1) \\ \mathsf{Doctor}(d_2) \\ \mathsf{Consultant}(c_1) \\ \mathsf{Doctor}(d_1) \\ \mathsf{Doctor}(c_1) \end{array} \right. \\ \right. \\ \right. \\ \left. \begin{array}{l} \mathsf{Consultant}(c_1) \\ \mathsf{Consultant}(c_1) \end{array} \right\} \\ \mathsf{Consultant}(c_1) \\ \mathsf{Consultant}(c_1) \\ \mathsf{Consultant}(c_1) \end{array} \right\} \\ \mathsf{Consultant}(c_1) \\ \mathsf{Consultant}(c_1$







 $\mathcal{O} \left\{ \begin{array}{l} \mathsf{Doctor} \equiv \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consulatant} \sqsubseteq \mathsf{Doctor} \end{array} \right. \\ \mathsf{DB} \left\{ \begin{array}{l} \mathsf{treats}(d_1, p_1) \\ \mathsf{Patient}(p_1) \\ \mathsf{Doctor}(d_2) \\ \mathsf{Consultant}(c_1) \end{array} \right. \\ \mathsf{DB}' \right. \end{array} \right.$

$$\mathsf{DB'} \begin{cases} \frac{\mathsf{treats}(d_1, p_1)}{\mathsf{Patient}(p_1)} \\ \frac{\mathsf{Doctor}(d_2)}{\mathsf{Consultant}(c_1)} \\ \frac{\mathsf{Doctor}(d_1)}{\mathsf{Doctor}(c_1)} \end{cases}$$

$$Q_1 \quad Q(x) \leftarrow \mathsf{Doctor}(y)$$









 $\mathcal{O} \left\{ \begin{array}{l} \mathsf{Doctor} \equiv \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consulatant} \sqsubseteq \mathsf{Doctor} \end{array} \right. \\ \mathsf{DB} \left\{ \begin{array}{l} \mathsf{treats}(d_1, p_1) \\ \mathsf{Patient}(p_1) \\ \mathsf{Doctor}(d_2) \\ \mathsf{Consultant}(c_1) \end{array} \right. \\ \mathsf{DB}' \right. \end{array} \right.$

$$\mathcal{Q}_1 \quad Q(x) \leftarrow \mathsf{Doctor}(y)$$

 $\mathsf{DB'} \begin{cases} \frac{\mathsf{treats}(d_1, p_1)}{\mathsf{Patient}(p_1)} \\ \frac{\mathsf{Doctor}(d_2)}{\mathsf{Consultant}(c_1)} \\ \frac{\mathsf{Doctor}(d_1)}{\mathsf{Doctor}(c_1)} \end{cases}$

 $\rightsquigarrow \qquad \{d_2, d_1, c_1\}$







 $\mathcal{O} \left\{ \begin{array}{c} \mathsf{Doctor} \equiv \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consulatant} \sqsubseteq \mathsf{Doctor} \end{array} \right.$ $\mathsf{DB'} \begin{cases} \frac{\mathsf{treats}(d_1, p_1)}{\mathsf{Patient}(p_1)} \\ \frac{\mathsf{Doctor}(d_2)}{\mathsf{Consultant}(c_1)} \\ \frac{\mathsf{Doctor}(d_1)}{\mathsf{Doctor}(c_1)} \end{cases}$ $\mathsf{DB} \begin{cases} \frac{\mathsf{treats}(d_1, p_1)}{\mathsf{Patient}(p_1)} \\ \frac{\mathsf{Doctor}(d_2)}{\mathsf{Consultant}(c_1)} \end{cases}$ $\rightsquigarrow \{d_2, d_1, c_1\}$ $Q_1 \quad Q(x) \leftarrow \mathsf{Doctor}(y)$ $\mathcal{Q}_2 \quad Q(x) \leftarrow \mathsf{treats}(x, y) \land \mathsf{Patient}(y)$









 $\mathcal{O} \left\{ \begin{array}{c} \mathsf{Doctor} \equiv \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consulatant} \sqsubseteq \mathsf{Doctor} \end{array} \right.$ $\mathsf{DB'} \begin{cases} \frac{\mathsf{treats}(d_1, p_1)}{\mathsf{Patient}(p_1)} \\ \frac{\mathsf{Doctor}(d_2)}{\mathsf{Consultant}(c_1)} \\ \frac{\mathsf{Doctor}(d_1)}{\mathsf{Doctor}(c_1)} \end{cases}$ $\mathsf{DB} \begin{cases} \frac{\mathsf{treats}(d_1, p_1)}{\mathsf{Patient}(p_1)} \\ \frac{\mathsf{Doctor}(d_2)}{\mathsf{Consultant}(c_1)} \end{cases}$ $\rightsquigarrow \qquad \{d_2, d_1, c_1\}$ $Q_1 \quad Q(x) \leftarrow \mathsf{Doctor}(y)$ $\mathcal{Q}_2 \quad Q(x) \leftarrow \mathsf{treats}(x, y) \land \mathsf{Patient}(y) \quad \rightsquigarrow \quad \{d_1\}$ Information Systems Group DEPARTMENT OF









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









Measuring and Repairing Incompleteness

• Use ontology \mathcal{O} (and query \mathcal{Q}) to generate a test suite









Measuring and Repairing Incompleteness

- Use ontology \mathcal{O} (and query \mathcal{Q}) to generate a test suite
- A test suite for ${\cal O}$ is a pair ${f S}=\langle {f S}_{\perp},{f S}_Q
 angle$
 - \mathbf{S}_{\perp} a set of ABoxes that are unsatisfiable w.r.t. $\mathcal O$
 - S_Q a set of paris $\langle A, Y \rangle$ with A an ABox and Y a query









Measuring and Repairing Incompleteness

- Use ontology \mathcal{O} (and query \mathcal{Q}) to generate a test suite
- A test suite for ${\cal O}$ is a pair ${f S}=\langle {f S}_{\perp},{f S}_Q
 angle$
 - \mathbf{S}_{\perp} a set of ABoxes that are unsatisfiable w.r.t. $\mathcal O$
 - S_Q a set of paris $\langle A, Y \rangle$ with A an ABox and Y a query
- A reasoner \mathcal{R} passes 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.









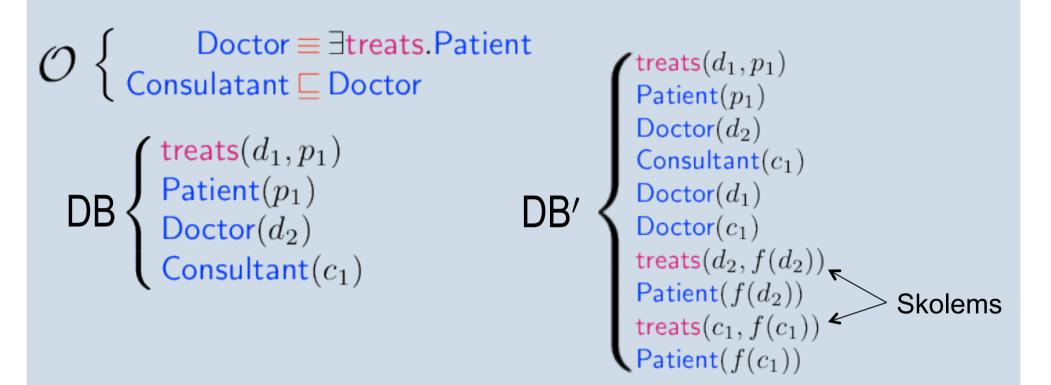
```
\mathcal{O} \left\{ \begin{array}{c} \mathsf{Doctor} \equiv \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consulatant} \sqsubseteq \mathsf{Doctor} \end{array} \right.
```

```
\mathsf{DB} \begin{cases} \frac{\mathsf{treats}(d_1, p_1)}{\mathsf{Patient}(p_1)} \\ \frac{\mathsf{Doctor}(d_2)}{\mathsf{Consultant}(c_1)} \end{cases}
```







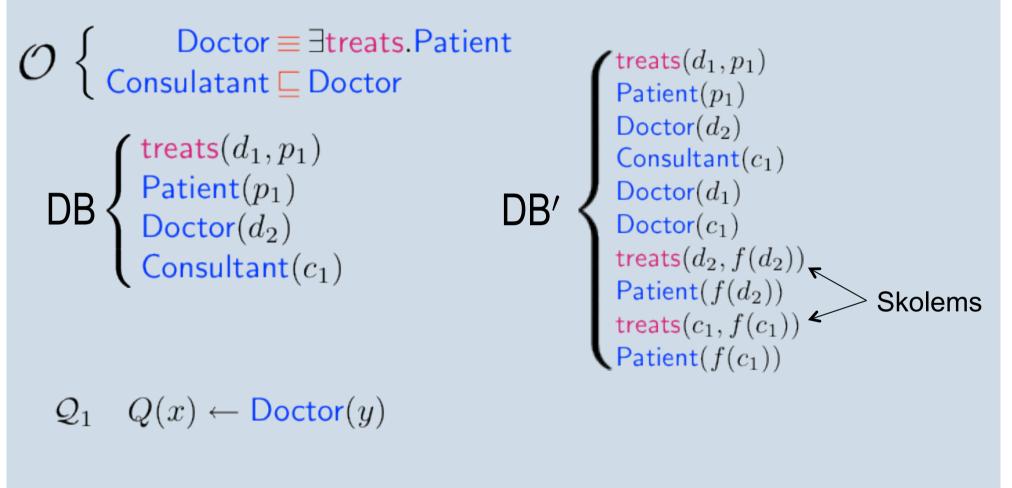












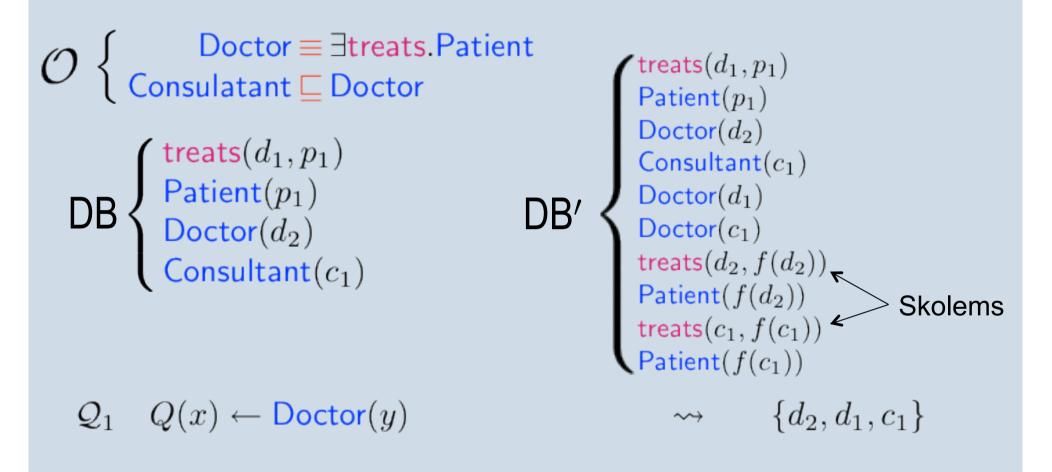
$$Q_1 \quad Q(x) \leftarrow \mathsf{Doctor}(y)$$









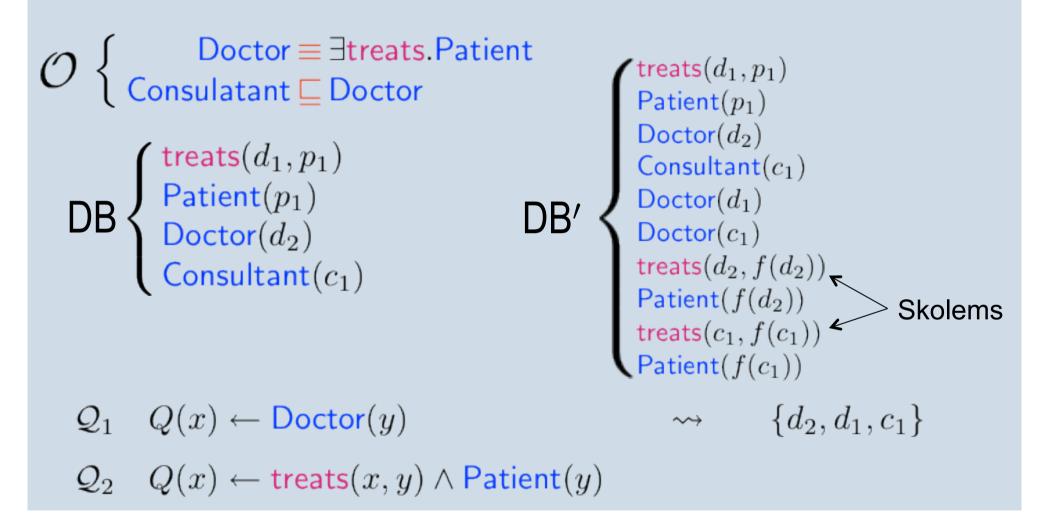










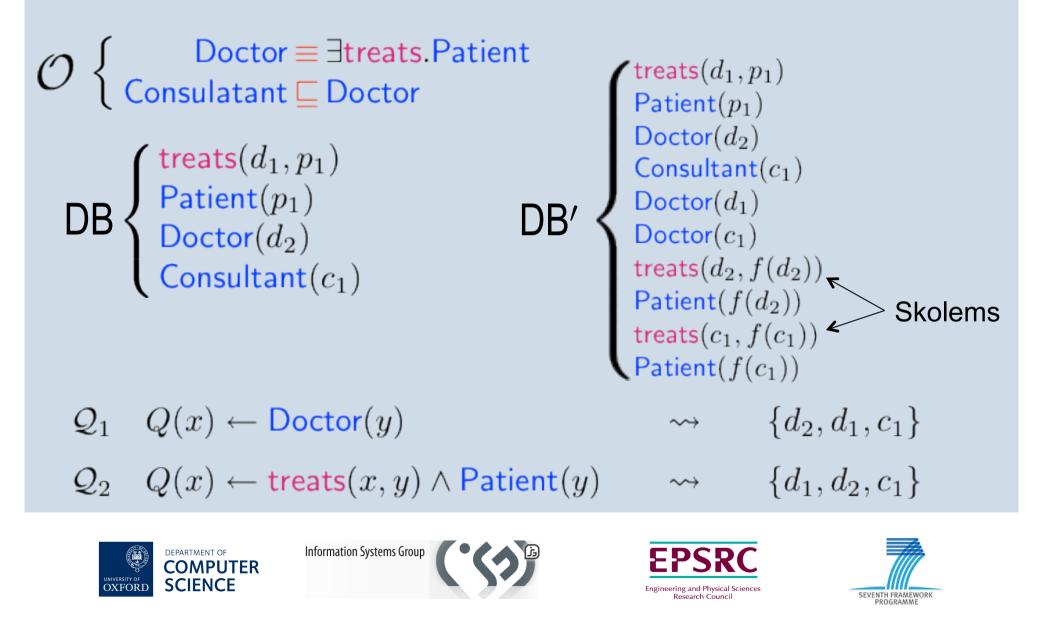












- RL reasoning w.r.t. OWL ontology ${\cal O}$ gives lower bound answer L









- RL reasoning w.r.t. OWL ontology ${\cal O}$ gives lower bound answer L
- Transform \mathcal{O} into strictly stronger OWL RL ontology
 - Transform ontology into $\mathsf{Datalog}^{\pm, \nu}$ rules
 - Eliminate V by transforming to Λ
 - Eliminate existentials by replacing with Skolem constants
 - Discard rules with empty heads
 - Transform rules into OWL 2 RL ontology O'









 RL reasonting w.r.t. O' gives (complete but unsound) upper bound answer U









Computing Upper Bound — Example

```
\mathcal{O} \left\{ \begin{array}{c} \mathsf{Doctor} \equiv \exists \mathsf{treats}.\mathsf{Patient} \\ \mathsf{Consulatant} \sqsubseteq \mathsf{Doctor} \end{array} \right.
```

 $\mathsf{DB} \begin{cases} \frac{\mathsf{treats}(d_1, p_1)}{\mathsf{Patient}(p_1)} \\ \frac{\mathsf{Doctor}(d_2)}{\mathsf{Consultant}(c_1)} \end{cases}$









Computing Upper Bound — Example

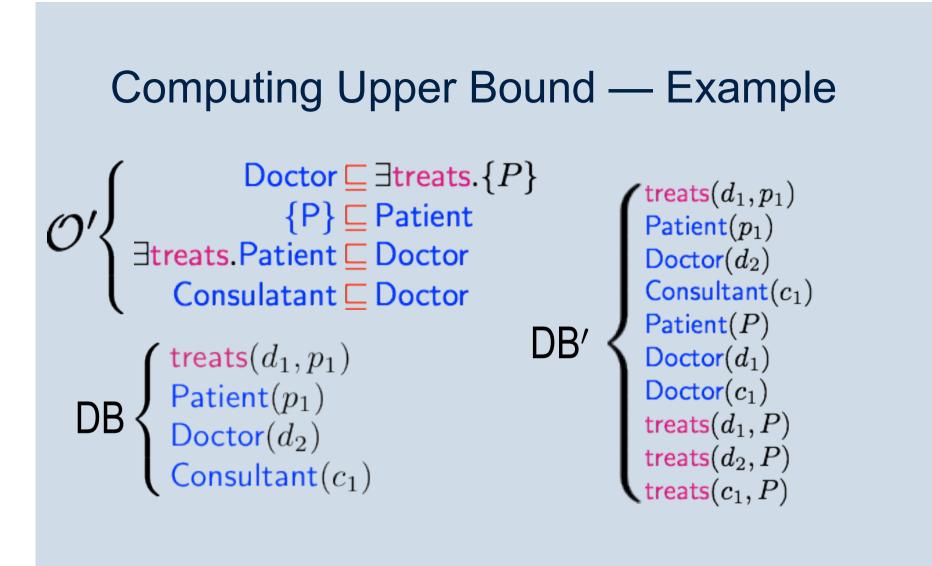
```
\mathcal{O}' \begin{cases} \mathsf{Doctor} \sqsubseteq \exists \mathsf{treats.} \{P\} \\ \{P\} \sqsubseteq \mathsf{Patient} \\ \exists \mathsf{treats.Patient} \sqsubseteq \mathsf{Doctor} \\ \mathsf{Consulatant} \sqsubseteq \mathsf{Doctor} \\ \mathsf{Consulatant} \sqsubseteq \mathsf{Doctor} \\ \mathsf{DB} \begin{cases} \mathsf{treats}(d_1, p_1) \\ \mathsf{Patient}(p_1) \\ \mathsf{Doctor}(d_2) \\ \mathsf{Consultant}(c_1) \end{cases}
```









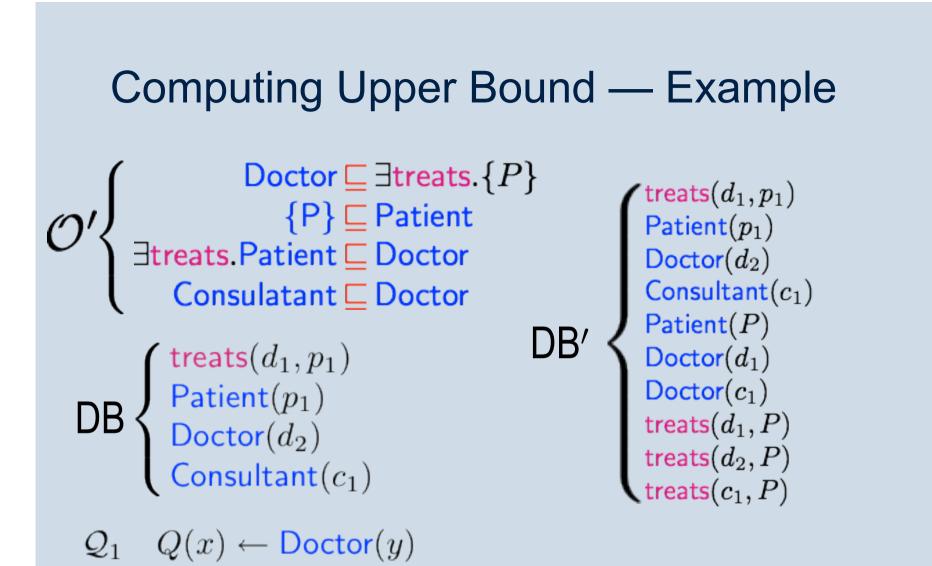












 $Q_1 \quad Q(x) \leftarrow \mathsf{Doctor}(y)$

DEPARTMENT OF COMPUTER SCIENCE

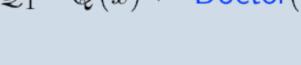
OXFORD







 $\mathcal{O}' \begin{cases} Doctor \subseteq \exists treats. \{P\} \\ \{P\} \subseteq Patient \\ \exists treats. Patient \subseteq Doctor \\ Consulatant \subseteq Doctor \\ doctor \\$ $\rightsquigarrow \{d_2, d_1, c_1\}$ $Q_1 \quad Q(x) \leftarrow \mathsf{Doctor}(y)$



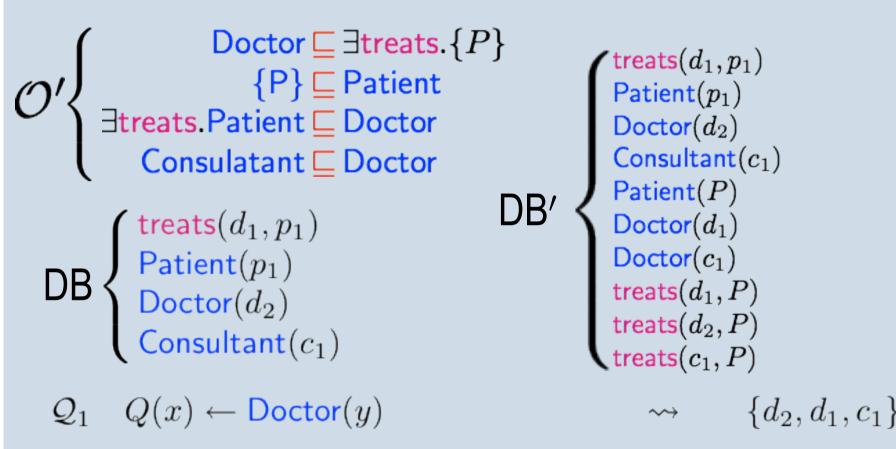








Computing Upper Bound — Example



 $Q_2 \quad Q(x) \leftarrow \mathsf{treats}(x, y) \land \mathsf{Patient}(y)$

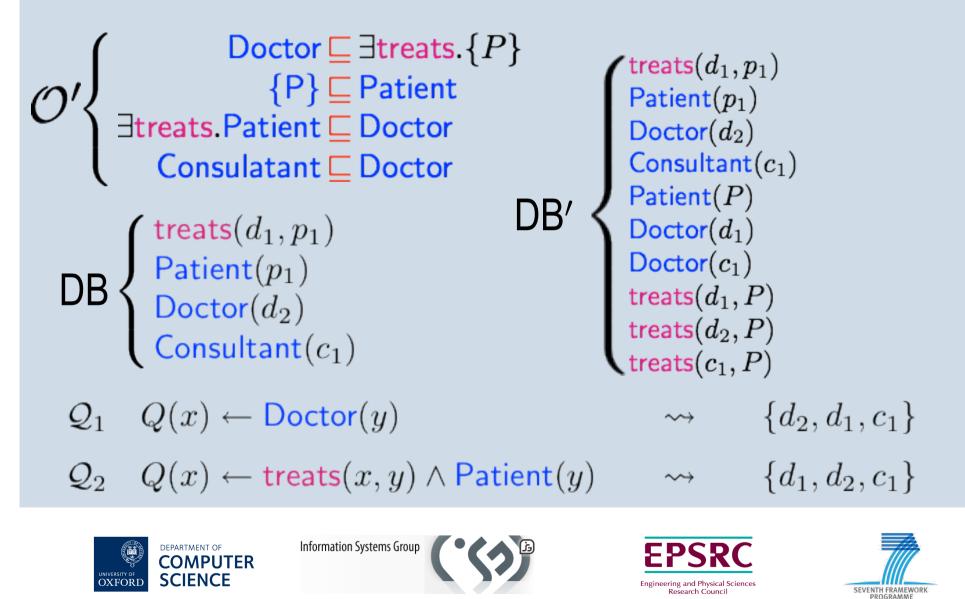








Computing Upper Bound — Example



Research Council

- RL reasonting w.r.t. 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 \ 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.









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.
- Parameterised complexity, new query evaluation techniques, etc.









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.
- Parameterised complexity, new query evaluation techniques, etc.

Consider progress on schema reasoning:

Year	$\mathcal{O} ext{-size}$	Complete	Time (s)
1995	$3,\!000$	No	10^{5}
1998	$3,\!000$	Yes	300
2005	30,000	Yes	30
2010	400,000	Yes	5









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.
- Parameterised complexity, new query evaluation techniques, etc.

Consider progress on schema reasoning:

Looking forward to similar progress on query answering!









Numerous exciting developments & research areas

- Rewriting: optimisations, extensions (datalog engines), etc.
- Materialisation: chase, repair, truth maintenance, upper bounds etc.
- Hybrid techniques (materialisation+rewriting), Datalog

Semantics \sqcap Scalability $\not\models \perp$!

Consider progress on schema reasoning:









Acknowledgements





Engineering and Physical Sciences Research Council











References

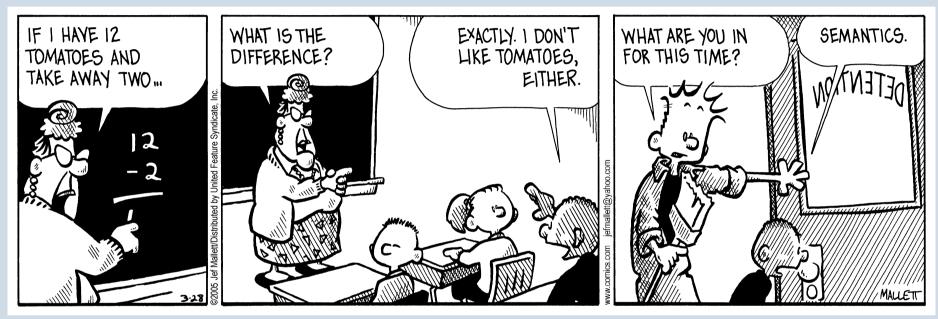
- [1] Armas Romero, Cuenca Grau, and Horrocks. Modular Combination of Reasoners for Ontology Classification. In Proc. of ISWC 2012 (to appear).
- [2] Kontchakov, Lutz, Toman, Wolter, Zakharyaschev: The Combined Approach to Ontology-Based Data Access. IJCAI 2011.
- [3] Stefanoni, Motik, Horrocks: Small Datalog Query Rewritings for EL. DL 2012
- [4] Kazakov, Kroetzsch, Simancik: Practical Reasoning with Nominals in the EL Family of Description Logics. KR 2012
- [5] Rodriguez-Muro, Calvanese: High Performance Query Answering over DL-Lite Ontologies. KR 2012
- [6] Motik, Horrocks, and Kim. Delta-reasoner: a semantic web reasoner for an intelligent mobile platform. In Proc. of WWW 2012.
- [7] Cuenca Grau, Motik, Stoilos, and Horrocks. Completeness Guarantees for Incomplete Ontology Reasoners: Theory and Practice. JAIR, 43:419-476, 2012
- [8] Cuenca Grau et al. Acyclicity Conditions and their Application to Query Answering in Description Logics. In Proc. of KR 2012.
- [9] Zhou, Cuenca Grau, and Horrocks. Efficient Upper Bound Computation of Query Answers in Expressive Description Logics. In Proc. of DL 2012







Thank you for listening



FRAZZ: © Jeff Mallett/Dist. by United Feature Syndicate, Inc.

Any questions?







