Summary

Semantic Web Technologies Lecture 9: SWT for the Life Sciences 2: Successes and challenges for ontologies

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Successes Exploiting the classification reasoning services Scalable querying of ontologies and data

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Representation Reasoning issues

- Only if Berners-Lee's vision of the SemWeb (as in the SciAm 2001 paper) has been realised?
- Absolute measures? e.g.,
 - User's (browsing and buying) usage of Amazon's recommender system with and without SWT
 - Information retrieval: compare precision and recall between a statistics-based and a SWT-based implementation of a
 - document system
 - Feasibility and performance of a set of user queries posed to a RDBMS and its RDF-ised version
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Examples in different application areas, using different features

• Data integration

- Instance classification (example today)
- Matchmaking and services
- Querying, information retrieval
 - Ontology-Based Data Access (completoday)
 - Aid in browsing large ontologies
 - Ontologies to improve NLP (more tomorrow)

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- Challenge: solution to problem y not possible yet (or very difficult to achieve) with current SWT, but in theory is (expected to be) feasible
- Failure: technology x claims to solve problem y but it does not and will not do so, or technology x is developed for a non-existing problem but does not solve real problems
 - Is y one that, at least in theory, can be solved with SWT?
 - Was y described too broadly, so that it solves only a subset of the cases?
 - Were there perhaps additional requirements put on a solution?
- Are disconnected technologies with ad-hoc patches a challenge to solve or a failure in devising a generic suite?
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- Offer and demand, perceptions, perspectives, expectations

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Instance classification with protein phosphatases (Wolstencroft et al, 2007)

- The setting:
 - Lots of sequence data in data silos that needs to be enriched with biological knowledge
 - Need to organise and classify genes and proteins into functional groups to compare typical properties across species
- The problems:
 - There is no proper, real life, use case that demonstrates the benefits of DL reasoning services such as taxonomic and instance classification
 - Limitations of traditional similarity methods, and automated protein motif and domain matching
 - Automation of p-domain analysis, but not for its interpretation (i..e, detects presence but not consequences for sub-family membership)

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- Maybe OWL reasoning can help with the interpretation of the analysis results:
 - That it does the classification of the (family of) proteins as good as a human expert for organisms *x* (in casu, human)
 - That the approach is 'transportable' to classification of the (family of) proteins in another organism of which much less is known (in casu, *Aspergillus fumigatus*), hence make predictions for those instances by means of classifying them
- Use taxonomic classification and instance classification reasoning services

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How it can be done

· Develop ontology for the subject domain, in OWL

- Extract knowledge from peer-reviewed literature
- Protein phosphatases; e.g.
 Class R5Phosphatase Complete

 (Protein and
 (hasDomain two TyrosinePhosphataseCatalyticDomain) and
 (hasDomain some TransmembraneDomain) and
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- Obtain instance data
 - Process protein sequences by InterProScan
 - Transform into OWL
- Put it together in some system with a reasoner
 - InstanceStore
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- Human phosphatases:
 - The reasoner as good as human expert classification
 - Identification of additional p-domains, refined the classification into further subtypes
- A. fumigatus phosphatates:
 - Some phosphatases did not fit in any class, representing differences between the human and A. fumigatus protein families
 - Identification of a novel type of calcineurin phosphatase (has extra domain, like in other pathogenic fungi)
- Overall: demonstration that ontology-based approach with automated reasoning has some advantages over (in addition to the) existing technologies & human labour, and resulted in discovery of novel biological information

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Web-based, graphical, ontology-based querying of lots of data (Calvanese et al, 2010)

- The setting:
 - Large amounts of data available on the Web, which can be accessed by canned or precomputed queries presented via web forms, or SQL
 - Domain expert wants more flexibility in data analysis and hypothesis testing, and independence from the sysadmin to do the queries for them

• The problems:

- There is no proper, real life, use case that demonstrates the benefits of scalable, user-usable, Ontology-Based Data Access
- That one has to know how the data is stored, instead of concerning oneself with what kind of information is in the database
- Domain expert-unfriendly query mechanisms (SQL, SPARQL)

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- Ontology-Based Data Access, to achieve data access at the 'what-layer', i.e., adding a semantic layer to the database
- Web-based, like most other bioinformatics resources
- Graphical querying to make it usable by the domain expert
- Usage of, mainly, reasoning services for querying the ontology and the data

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How it can be done

- Develop ontology of the subject domain, in OWL
 - Reverse engineering existing database HGT-DB (http://genomes.urv.cat/HGT-DB/), further manual improvements to create a proper conceptual data model
 - Simplify this conceptual data model into the appropriate OWL language (*DL-Lite_A*, which is roughly OWL 2 QL)
- Create mappings between the terms in the ontology to SQL queries over the database
 - Using the OBDA Plugin for Protégé
 - Oracle database (can also be PostgreSQL, DB2, ...), 4GB genomics database (HGT-DB), tables with 16-46 columns
- Connect this to an OBDA-enabled reasoner
 - In this case: QUONTO (but can be others)

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Example: Diagram – DL-lite_A correspondence



Figure 2: Section of the HGT application ontology.

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Summary

Formalisation of the graphical elements

Element name	Graphical Representation	Semantics of ontology elements	Semantics of building blocks of query graphs
Class node	С	C	C(x)
	C, D		C(x), D(x)
Is-a link		$C \sqsubseteq D$	
Attribute node and link	C A	$\delta(A) \sqsubseteq C$ $\rho(A) \sqsubseteq \top_d$	C(x), A(x,y)
Role link	C P D	$\exists P \sqsubseteq C \\ \exists P^- \sqsubseteq D$	C(x), R(x,y), D(y)

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Example: mapping concepts & relations of the Ontology to SQL query over the relational database

SELECT id, abbrev	$\sim \rightarrow$	OrganismHasGene(
FROM organism		gene(<i>id</i>),
JOIN genes		$\operatorname{organism}(abbrev))$
ON abbrev = idorganism		
SELECT id, kegg	\sim	GeneHasGeneFunction(
FROM genes		gene(id), function(id))
		KEGG(function(id), kegg)

Figure 3: Extract of the mapping from the HGT-DB database to the $DL-Lite_A$ application ontology.

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Queries

- SPARQL queries for conjunctions and equalities
- Epistemic queries in *EQL-Lite* for constraints involving inequalities and string matching
 - Imposes constraints on top of the certain answers retrieved by a *DL-Lite_A* conjunctive query
 - Result obtained by:

i. computing the certain answers for the CQ $q(\vec{y}) \leftarrow conj(\vec{z})$ (with $conj(\vec{z})$ the conjunction of atoms, and \vec{y} a vector comprising the variables in \vec{x} and in \vec{w}),

ii. filtering the resulting tuples according to the constraint expression $cons(\vec{w})$, and

iii. projecting onto \vec{x} (a vector comprising the variables corresponding to the highlighted nodes in the query pane)

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- Demo of the WONDER system (Web-ONtology baseD Extraction of Relational data)
 - Builds upon the theory, technology, and implementation developed for Ontology-Based Data Access
 - Graphical ontology browsing, query formulation, and query execution in a Web browser
 - Rigorous formal characterisation and uses a coupling with an OWL file
 - (U)CQs (in SPARQL syntax) and EQL-Lite queries managed by the DIG-QUONTO reasoner
- Performance good, GUI insignificant influence on performance
- Usability testing: usable, and domain experts came up with a range of new queries to analyse the data

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- Performance good, GUI insignificant influence on performance
- Usability testing: usable, and domain experts came up with a range of new queries to analyse the data

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Summary

Additional features

- WONDER currently focuses on querying one database
- OBDA architecture allows for querying incomplete data (data integration scenario¹)
- Querying of the application ontology itself, as well as a combination of querying the ontology and the data²
 - in certain settings, possible to include queries that use the knowledge in the ontology for which there is no data in the database, and still retrieve the right results

¹A. Amoroso, G. Esposito, D. Lembo, P. Urbano, and R. Vertucci. Ontology-based data integration with MASTRO-I for configuration and data management at SELEX Sistemi Integrati. In Proc. of SEBD 2008, pages 8192, 2008.

 $^{^2}$ C. M. Keet, R. Alberts, A. Gerber, and G. Chimamiwa. Enhancing web portals with Ontology-Based Data Access: the case study of South Africa's Accessibility Portal for people with disabilities. In Proc. of OWLED 2008, 2008. CEUR-WS Vol 432

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Informal overview of kind of knowledge in ADOLENA



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Summary

Sample query in OBDA Plugin

Properties 🖌 🔶 Individuals 👘 🚍 Forms 🕴 👭	Datasource Manager 🛛 🥵 ABox Queries			
QUERY EDITOR				
Query name: Unamed Description: No descriptio	on			
Ouery (SPAROL):				
SELECT \$ nome \$ description WHEPE \$ \$	endfitume 'NA PrAccistive Device' \$x			
select share subscription where $\{\phi\}$	imbMobility/ \$x :description \$description			
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ax mame aname t				
	Advanced properties Execute Save			
▲ ▼	Advanced properties Execute Save			
A	Advanced properties Execute Save			
A 'CE Power'	Advanced properties Execute Save B 'Portabile & Economic motorised wheelchair.'			
A 'CE Power' 'Dassie 200 M'	Advanced properties Execute Save B 'Portabile & Economic motorised wheelchair.' 'The versatile and nimble Dassie 200 M is the perfec			
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A 'CE Power' 'Dassie 200 M' 'Gazelle 250 S' 'Predator 4 x 4' 'Roman power wheelchair' 'Karma Power'	Advanced properties Execute Save Portabile & Economic motorised wheelchair.' The versatile and nimble Dassie 200 M is the perfec Slick and stylish would be the best way to describe t The mean beast of wheelchairs, yep it is the ALL TE Wheelchair motorised' 'Karma KP40 wheelchair, powerfull fixed base wheel			
A 'CE Power' 'Dassie 200 M' 'Gazelle 250 S' 'Predator 4 x 4' 'Roman power wheelchair' 'Karma Power' 'Karma Power' 'Battery powered mobility vehicle'	Advanced properties Execute Save B 'Portabile & Economic motorised wheelchair.' 'The versatile and nimble Dassie 200 M is the perfec 'Slick and stylish would be the best way to describe t 'Wheelchair motorised' 'Karma KP40 wheelchair, powerfull fixed base wheel 'We supply battery powered mobility vehicles. They			

q(x) :- Device(x), assistsWith(x, y), UpperLimbMobility(y)

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Challenges Representation Reasoning issues

Challenges

- RDF triple stores vs. RDBMSs vs OWL ABoxes in memory; more generally:
 - Making 'legacy' (operational) systems 'Semantic Web compliant'
 - Add a 'wrapper' over the legacy system so that from the outside it looks like it uses SWT
- How to integrate rules other than at instance level
- Modularization
- Semantics-based language transformations
- Coordination among tools with different functionalities

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Language limitations considerations

- Known trade-offs between expressiveness and computational complexity
- Different ontology developers and their scopes (and purposes of the ontologies):
 - to some, there is more in OWL/OWL2 than needed and used (recollect slide 32 of lecture 8)
 - to some, there is not enough (some of the limitations and extensions discussed in lecture 2, 6 and 7)
- From a logician's perspective, language limitations are not failures per sé, only *challenges* to find the more interesting and useful combinations of features
- From a modeller's perspective, the trade-offs can be such that it is deemed a *failure* with respect to the expectations and application needs
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- *n*-ary relations, where n > 2
- "Hepatitis hasSymptom Fever in most but not all cases"
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 - {ψ | ψ}|, u| as "generally, if an object belongs to ψ, then it belongs to ψ, with a probability in [l, u]"
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Limitations as identified by users/modellers (Schulz et al, 2009)

• ... Diabetes example continued

- Authors' proposal to put it in the ABox with arithmetic operators, e.g. "

 <u>|DiabeticHuman|</u> = 0.028"
- Yet another: represent the *probability* of a human having diabetes mellitus
- What are the pros and cons of each option w.r.t. subject domain semantics, Ontology, and the ontology languages?
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Limitations as identified by users/modellers (Schulz et al, 2009)

- "Concussion of the brain without loss of consciousness", and the temporal aspects (recollect lecture 6)
- "aspirin prevents myocardial infarction"
 - Let us assume that is total prevention (though we could add a probability to it)
 - This only holds for humans actually ingesting aspirin, not for the substance itself
 - It then intends to say that the human taking aspirin will not have a myocardial infarction at all times in the future, which can be represented in a suitable temporal logic with the CP.

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- The standard reasoning services (recollect lecture 5) are obviously sorted out
- Performance issues for the 'debugging' and explanation reasoning, and how to provide the 'best' explanation
- Querying OWL 2 DL, and any ABox data
- Additional reasoning scenarios

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Challenges

Scenarios

- 1. Supporting the ontology development process
- 2. Classification
- 3. Model checking (violation)
- 4. Finding gaps in an ontology & discovering new relations
 - Deriving types and relations from instance-level data
 - · Computing derived relations at the type level
- 5. Comparison of two ontologies ([logical] theories)
- 6. Reasoning with part-whole relations
- 7. Using (including finding inconsistencies in) a hierarchy of relations
- 8. Reasoning across linked ontologies
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Checking against instances

Usual model checking

- Model checking against real instances in the ABox/Database
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 - To spot "redundant" DL-concepts w.r.t. the data-needs
- Model violation
 - Reducing the amount of instances to only those that do not violate the TBox (or: the more inconsistencies, the better)
 - For instance, to find a few candidate molecules that satisfy a given set of properties, out of a large pool of possibly suitable molecules; e.g., for drug discovery in pharmainformatics, tyre production

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Challenges

Checking against instances

- Usual model checking
- Model checking against *real* instances in the ABox/Database
 - For each DL-concept in the OWL-formalised ontology (representing a universal), there has to be at least one ABox instance (as representation of the entity in reality)
 - To spot "redundant" DL-concepts w.r.t. the data-needs
- Model violation
 - Reducing the amount of instances to only those that do not violate the TBox (or: the more inconsistencies, the better)
 - For instance, to find a few candidate molecules that satisfy a given set of properties, out of a large pool of possibly suitable molecules; e.g., for drug discovery in pharmainformatics, tyre production

Challenges

Discovering information

- The idea is that the combination of bio-ontologies, instances, and automated reasoning services somehow can find either the missing relations, or the types, or both
- How can one find what is, or may, not be in the ontology but ought to be there?
- At the TBox-level
 - computing derived relations (object properties)
 - find out where relations that are known by the developer have not yet been added to the ontology (finding 'known gaps')
 - add 'ontological' notions with top type 'whole' in a partonomy; e.g., 17 types of macrophage in the FMA each must be part of something
 - flag classes that have no relation (no or no is_a) to anything else in the ontology

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Challenges

Discovering information

- For the TBox through querying the data (ABox, RDBMS)
 - i. "for each x:X, y:Y, r:R, XRY, does there exist a z:Z, s:S, such that there exist ≥ 1 x and xsz?"
 - ii. "for each x:X, y:Y, *r*:*R*, X*R*Y, does there exist an x*s*z and an x*t*a where z:Z, *s*:*S*, a:A, *t*:*T* hold?"
 - iii. Find-me-anything-you-have: "for each x:X, return any $r_1, ... r_n$, their type of role and the concepts $Y_1, ... Y_n$ they are related to"



Introduction

Challenges

Summary



Introduction

Successes Exploiting the classification reasoning services Scalable querying of ontologies and data

Challenges Representation Reasoning issues