

C. Maria Keet

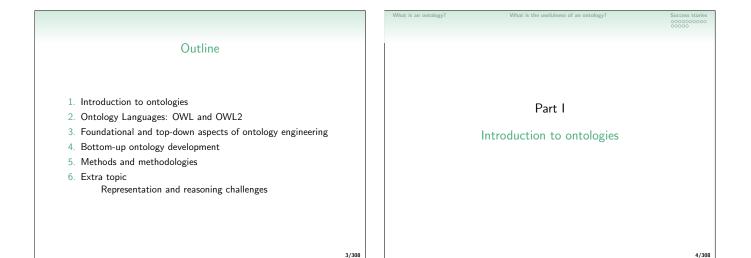
KRDB Research Center Free University of Bozen-Bolzano, Italy

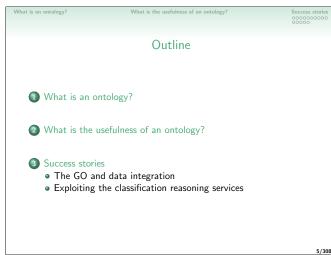
Masters Ontology Winter School 2010 KRR Group, Meraka Institute, South Africa, 26-30 July 2010

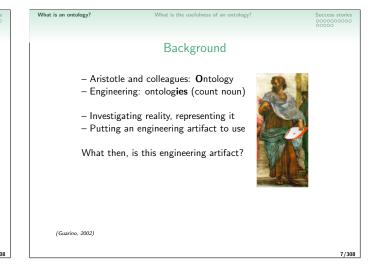
Housekeeping points

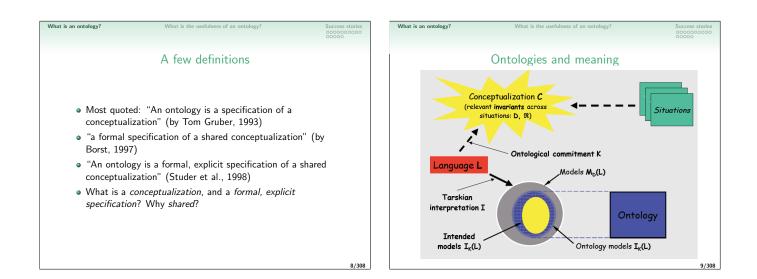
- This course consists of lectures and exercises
- Each lecture takes about 2.5 hours, labs 45 minutes
- Following the lectures will be easier when you have read the recommended reading beforehand and it is assumed the student is familiar with first order logic and conceptual data modelling, such as UML and ER
- The topics covered in this course are of an **introductory** nature and due to time constraints only a selection of core and elective topics will be addressed.
- These slides serve as a teaching aid, not as a neat summary
- Course webpage, with introduction, references, and schedule: http://www.meteck.org/teaching/SA/MOWS100ntoEngCouse.html

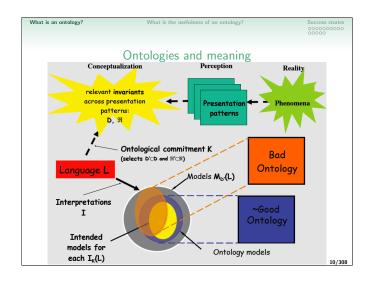
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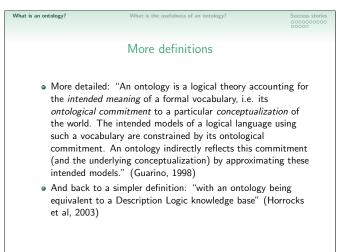


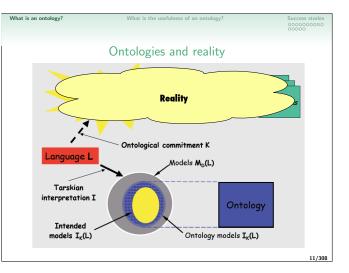


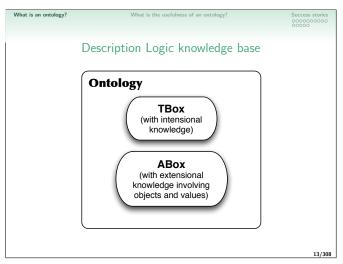


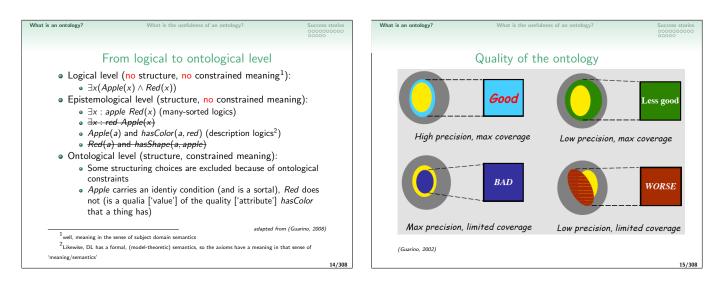


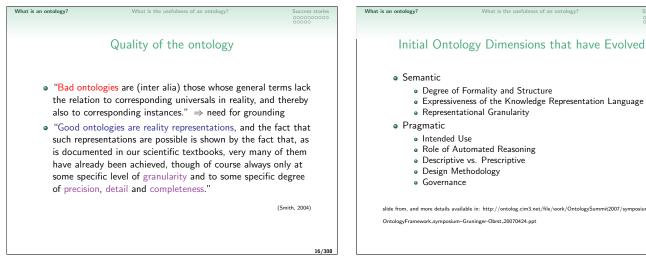


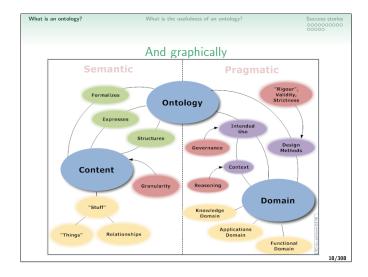


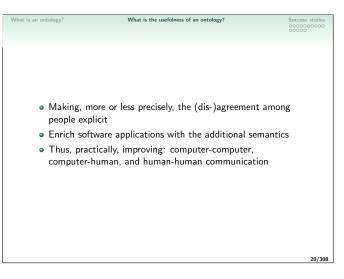












What is the usefulness of an ontology? Examples in different application areas, using different Success? features • Only if Berners-Lee's vision of the Semantic Web (as in the SciAm 2001 paper) has been realised? How much "semantics" (with ontologies)? • Data(base) integration (example today) SemWeb stack, technologies • Instance classification (example today) Absolute measures? e.g., Matchmaking and services Usage of Amazon's recommender system with and without ontologies Information retrieval: compare precision and recall between a • Querying, information retrieval statistics-based and a ontologies-mediated document system Ontology-Based Data Access · Feasibility and performance of a set of user queries posed to a RDBMS Ontologies to improve NLP and its RDF-ised version • Bringing more quality criteria into conceptual data modelling Relative measures to develop a better model (hence, a better quality software According to whom is it a success? system) philosopher, logician, engineer, domain expert, CEO... • What was taken as baseline material? e.g., • Orchestrating the components in semantic scientific . from string search in a digital library to ontology-annotated sorting workflows, e-learning, etc. of query answer · from no or clustering-based instance classification to one with OWL-based knowledge bases 23/308 21/308

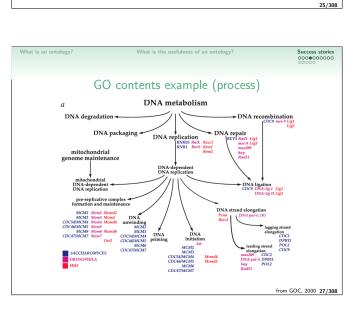
/hat is an ontology?	What is the usefulness of an ontology?	Success stories •••••••• •••••	What is an ontology?	What is the usefulness of an ontology?	Success st 0000000 00000
	Early bioinformatics			Scope and requirements	
 '80s-early'S Need to st Several 'model of the fruit zebrafish Compare g One of worm , inferre yeast, What (across) What annots organic 	in technologies to sequence genomes in the 90s, as well as more technologies for prote core the data: in databases ('90s) odel organism' databases with genes (and tfly, yeast, mouse, a flowering plant, flatwo genes and genomes bservation (of many): About 12% (some 18,0 genes encode proteins whose biological roles of do from their similarity to their (putative) orth comprising about 27% of the yeast genes (al else can we infer from comparing genes and g is species)? about the possibility of automated transfer of ations from the model organisms to less 'fancy isms based on gene and protein sequence similar improve human health or agriculture?	ins genomes) prm, 00) of the could be ologues in pout 5,700) renomes f biological /	transferring Methods for Main requir One new the gen <i>functior</i> To take termino UniProt Databas Organiz vastly d Any sys	eds a shared, controlled, vocabulary for ann the products, the location where they are act in they perform they be on board and be compatible with existing plogies, like gene and protein keyword datab t, GenBank, Pfam, ENZYME etc. se interoperability among, at least, the moc	notation of ive, the pases such as del organism knowledge at ponstantly

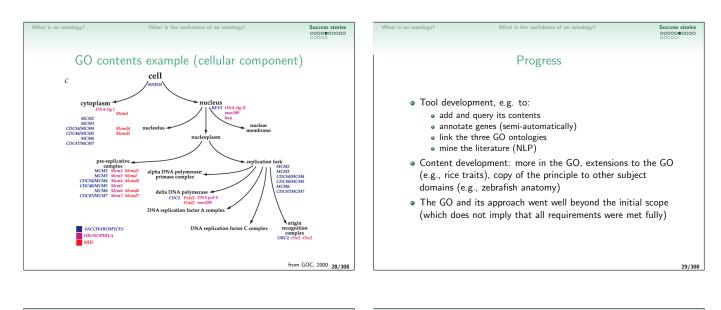
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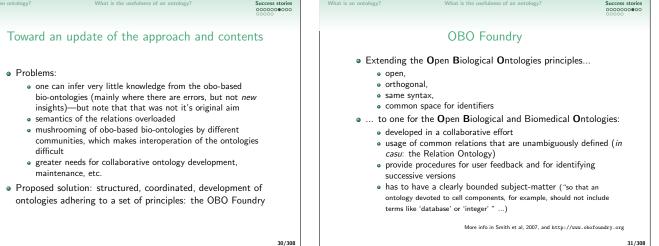
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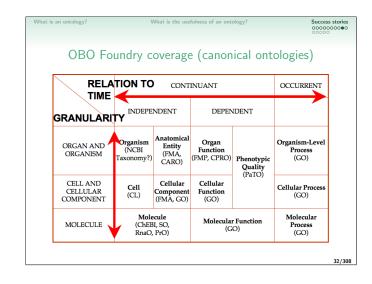
Success stories How to meet such requirements? • Two main strands in activities: Very early adopters from late 1990s (by sub-cellular bio), i.e., starting without Semantic Web Technologies Early adopters from mid 2000s (e.g., eco and agri), starting with Semantic Web Technologies • The Gene Ontology Consortium Initiated by fly, yeast and mouse database curators³ and others came on board (see http://ww.geneontology.org for a full list) In the beginning, there was the flat file format .obo to store the ontologies, definitions of terms and gene associations Several techniques on offer for data(base) integration that could be experimented with

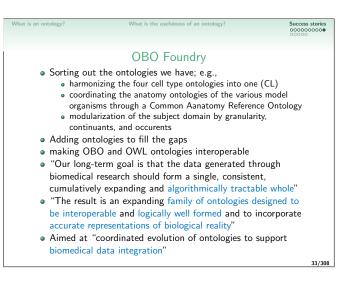
³more precisely: FlyBase (http://www.flybase.bio.indiana.edu), Berkeley Drosophila Genome Project (http://fruitfly.bdgp.berkeley.edu), Saccharomyces Genome Database (http://genome-www.stanford.edu), and Mouse Genome Database and Gene Expression Database (http://www.informatics.jax.org).



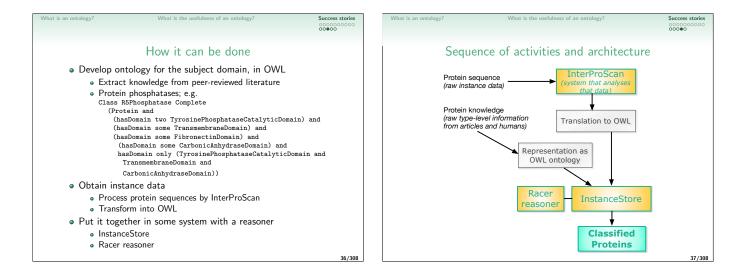


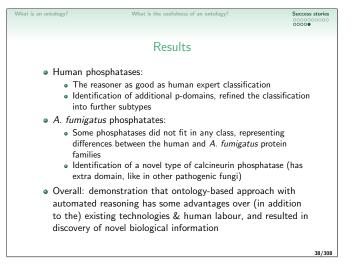


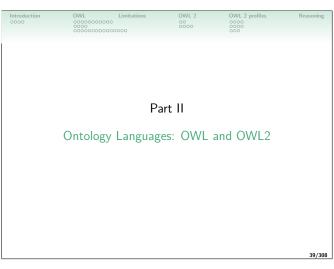




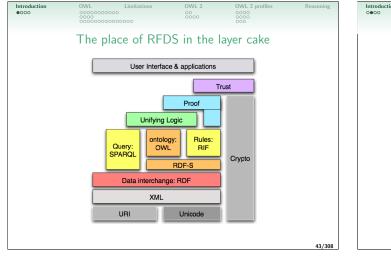
Success storie Instance classification with protein phosphatases (Wolstencroft Idea et al, 2007) • The setting: • Maybe OWL reasoning can help with the interpretation of the • Lots of sequence data in data silos that needs to be enriched analysis results: with biological knowledge • That it does the classification of the (family of) proteins as Need to organise and classify genes and proteins into good as a human expert for organisms x (in casu, human) functional groups to compare typical properties across species • That the approach is 'transportable' to classification of the • The problems: (family of) proteins in another organism of which much less is • There is no proper, real life, use case that demonstrates the known (in casu, Aspergillus fumigatus), hence make benefits of DL reasoning services such as taxonomic and predictions for those instances by means of classifying them instance classification • Use taxonomic classification and instance classification Limitations of traditional similarity methods, and automated reasoning services 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) 34/308 35/308

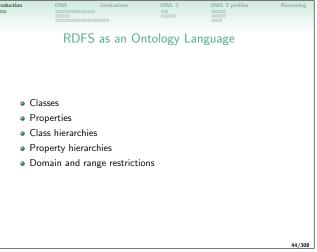


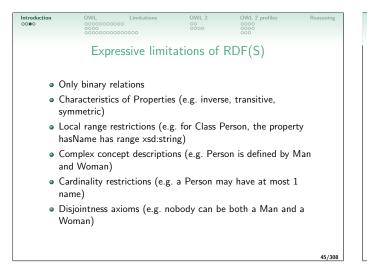


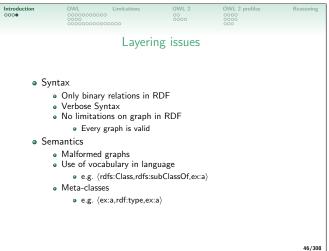


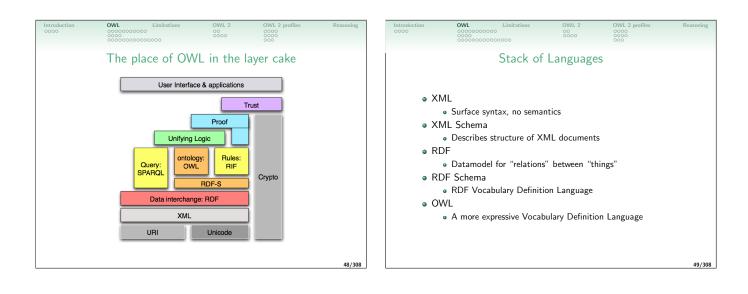
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-	Ou roduction Limitations of RDFS	ıtline				Toward one o	ntology la	nguage	
• (VL Design of OWL DWL and Description Logi DWL Syntaxes	CS			F- • La	ethora of ontology langu logic, DAML, OIL, DAM ck of a lingua franca; he	L+OIL, nce, ontology		
0 OW	nitations VL 2 OWL 2 DL				• Ac	bblems even on the synta dvances in expressive DL automated reasoners for CT++, then Racer)	languages ar		
• (• (VL 2 profiles OWL 2 EL OWL 2 QL OWL 2 RL					nitations of RDF(S) as S	Semantic We	b 'ontology language	,
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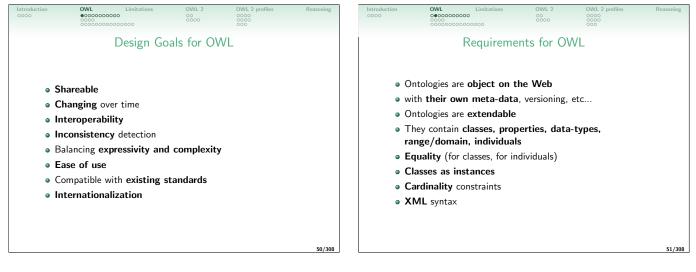


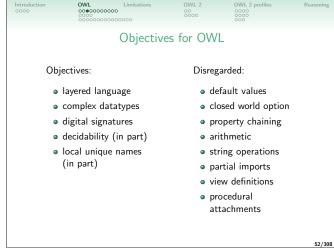


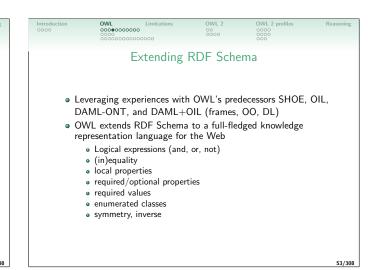


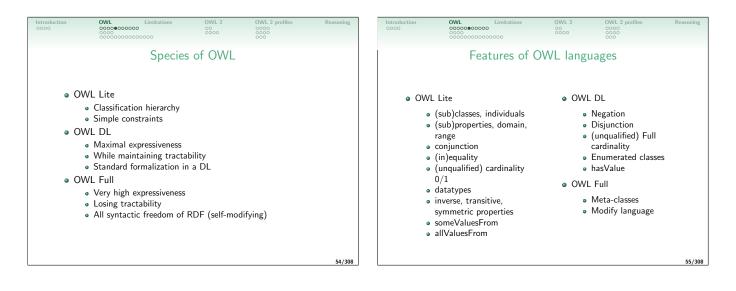


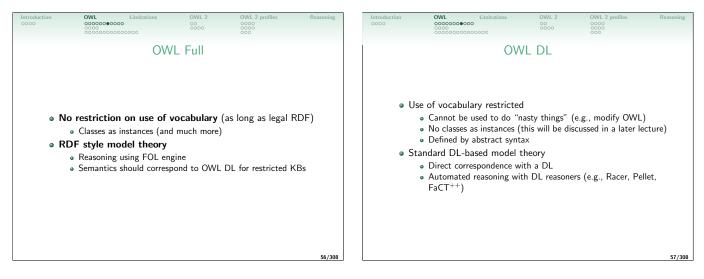


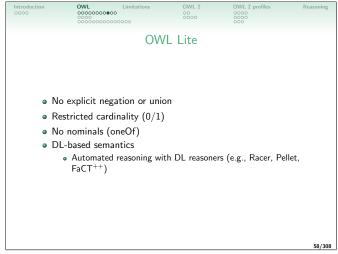


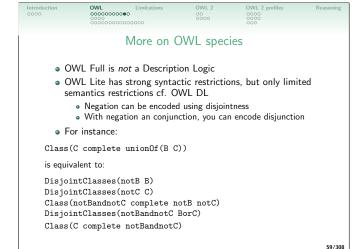


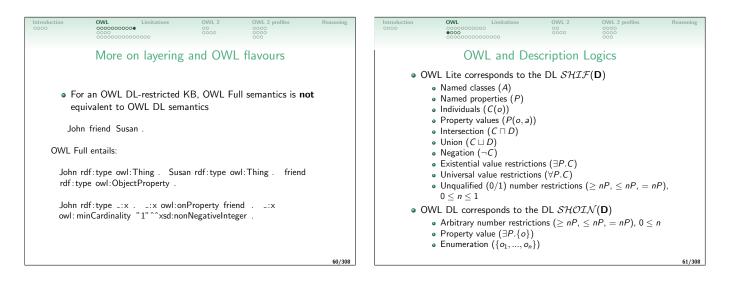






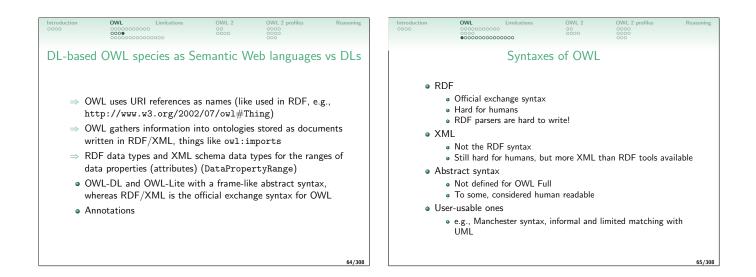


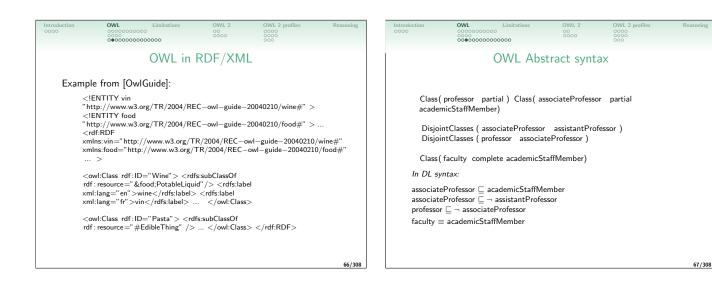




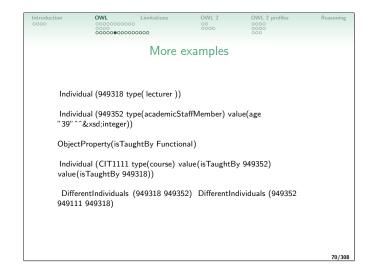
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OWL	Construct	DL	Example		
interse	ectionOf	$C_1 \sqcap \sqcap C_n$	Human 🗆	Male	
union	Df	$C_1 \sqcup \sqcup C_n$	Doctor ⊔	Lawyer	
compl	ementOf	$\neg C$	$\neg Male$		
oneOf		$\{o_1,, o_n\}$	{john, ma	ary}	
allVal	JesFrom	∀P.C	∀hasChild	1.Doctor	
some\	/aluesFrom	$\exists P.C$	∃hasChild	1.Lawyer	
value		∃ <i>P</i> .{ <i>o</i> }	∃citizenC	of.USA	
minCa	rdinality	$\geq nP.C$	$\geq 2hasCl$	nild.Lawyer	
maxC	ardinality	$\leq nP.C$	≤ 1 hasCl	nild.Male	
cordin	ality	= nP.C	= 1 has Pa	arent.Female	

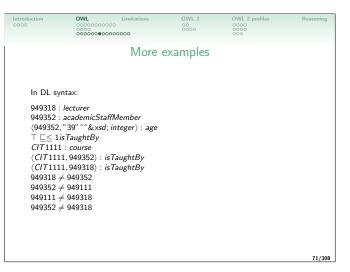
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		OWL a	ixioms		
OWL	Axiom	DL	Exai	mple	
SubCl	assOf	$C_1 \sqsubseteq C_2$	Hun	nan 🗆 Animal 🗆 l	Biped
Equiva	alentClasses	$C_1 \equiv \dots \equiv$	C _n Mar	$n \equiv Human \sqcap Ma$	le
SubPr	opertyOf	$P_1 \sqsubseteq P_2$	hasl	Daughter ⊑ hasC	hild
Equiva	alentProperties	$P_1 \equiv \dots \equiv$	P _n cost	\equiv price	
Samel	ndividual	$o_1 = =$	o _n Pres	sident_Bush = G	W_Bush
Disjoi	ntClasses	$C_i \sqsubseteq \neg C_i$	Mal	$e \sqsubseteq \neg Female$	
Differ	entIndividuals	$o_i \neq o_i$	john	$p \neq peter$	
invers	eOf	$P_1 \equiv P_2^-$	has($Child \equiv hasParen$	t-
Transi	tive	$P^+ \sqsubseteq \tilde{P}$	ance	$estor^+ \sqsubseteq anceston$	r
c	etric	$P \equiv P^{-}$	coni	$nectedTo \equiv conn$	ectedTo ⁻



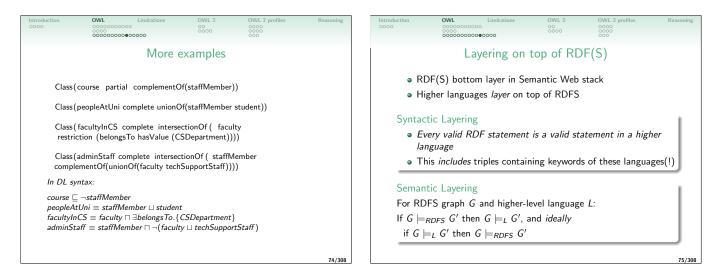


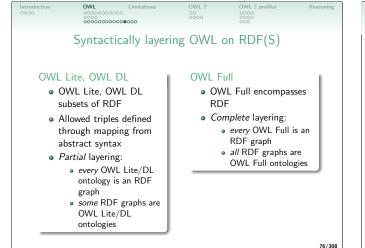
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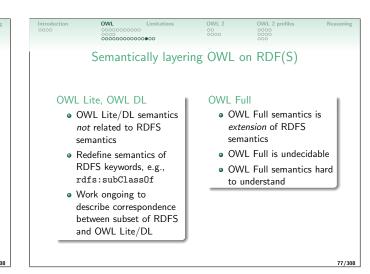




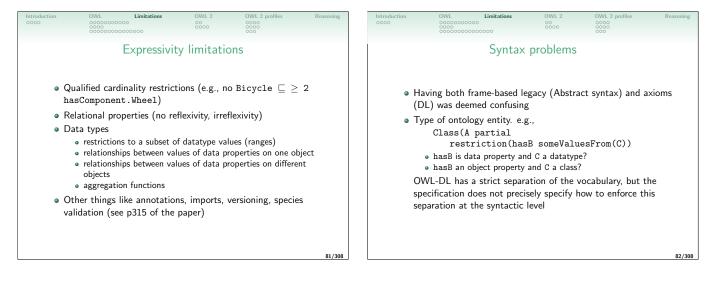
More examples More examples Class (firstYearCourse partial restriction (isTaughtBy allValuesFrom (Professor))) In DL syntax: Class (mathCourse partial restriction (isTaughtBy hasValue (949352))) In DL syntax: Class (academicStaffMember partial restriction (teaches someValuesFrom (undergraduateCourse))) In DL syntax: Class (course partial restriction (isTaughtBy minCardinality(1))) In State (StaffMember partial restriction (hasMember minCardinality(10))) Class (department partial restriction (hasMember minCardinality(10))) In State (StaffMember partial restriction (hasMember minCardinality(10)))	Introduction 0000	OWL Li	imitations 00	OWL 2 00 0000	OWL 2 profiles	Reasoning	Introduction 0000	OWL Limitations 0000000000 00000000000000000	OWL 2 00 0000	OWL 2 profiles	Reasoning
(Professor))) In DL syntax: Class (mathCourse partial restriction (isTaughtBy hasValue (949352))) In DL syntax: first YearCourse □ ∀isTaughtBy.Professor mathCourse □ ∃isTaughtBy.Professor (undergraduateCourse))) class (course partial restriction (isTaughtBy minCardinality (1))) Class (department partial restriction (hasMember minCardinality(10)) ln DL syntax:			More e	examples				More	examples		
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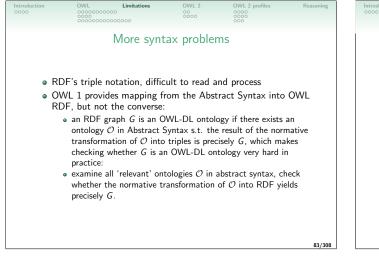


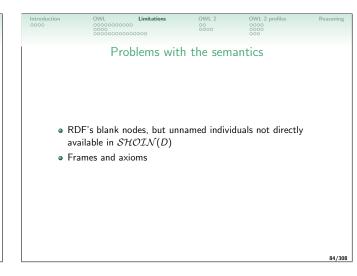




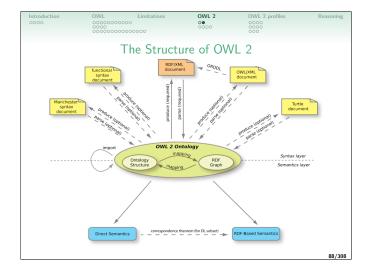
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		DL vs. RDF translation from Abstract	Syntax			OWL Lite/	DL vs. R	DF	
 Example Class(Hurr 	aan partial Animal restriction(hasLegs ca	rdinality(2)) IIIValuesFrom(xsd:string)))		• Exa	t every RD imple: rdf:type	F graph is OV	VL Lite/DL	ontology	
Human Human -:X1 -:X1 -:X1 Human -:X2 -:X2 -:X2 -:X2	rdf:type rdfs:subClassOf rdfs:subClassOf rdf:type owl:onProperty owl:cardinality rdfs:subClassOf rdf:type owl:onProperty owl:allValuesFrom	owl:Class Animal .:X1 owl:Restriction hasLegs "2"8sd:nonNegativeInteger .:X2 owl:Restriction hasName xsd:string			w to check Construct Translate If <i>G</i> = <i>G</i> ′,	whether an R	ology <i>O</i> in <i>i</i> h <i>G'</i> /L DL	is OWL DL? Abstract Syntax	
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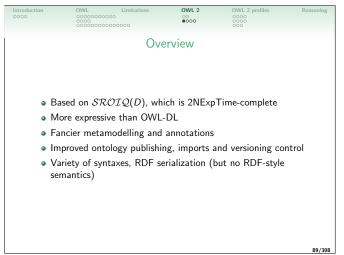


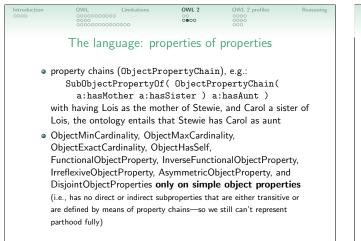


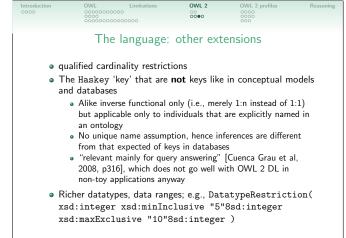


Introduction OWL Limitations OWL 2 OWL 2 profiles Reasoning 0000 000000000 00 0000 0000 0000 0000 0000 0000 0000 0000 0000	Introduction OWL Limitations OWL 2 OWL 2 profiles Reasonin 0000 000000000000000000000000000000000000
Aims	Some general points
 Address as much as possible of the identified problems (previous slides and "the next steps for OWL 2" paper) Task: compare this with the possible "future extensions" of the "the making of an ontology language" paper 	 OWL 2 a W3C recommendation since 27-10-'09 Any OWL 2 ontology can also be viewed as an RDF graph (The relationship between these two views is specified by the Mapping to RDF Graphs document) Direct, i.e. model-theoretic, semantics (⇒ OWL 2 DL) and an RDF-based semantics (⇒ OWL 2 full) Primary exchange syntax for OWL 2 is RDF/XML, others are optional Three profiles, which are sub-languages of OWL 2 (syntactic restrictions)
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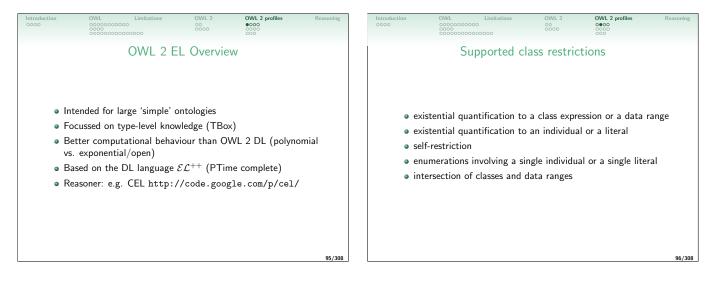


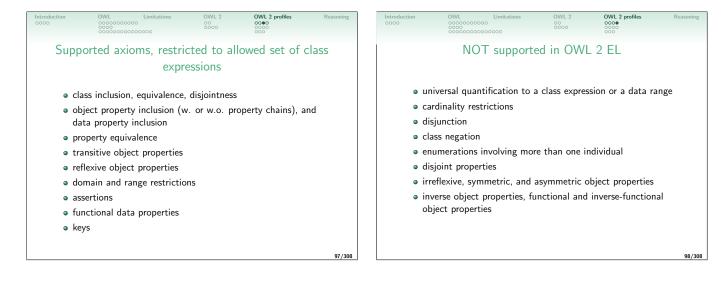


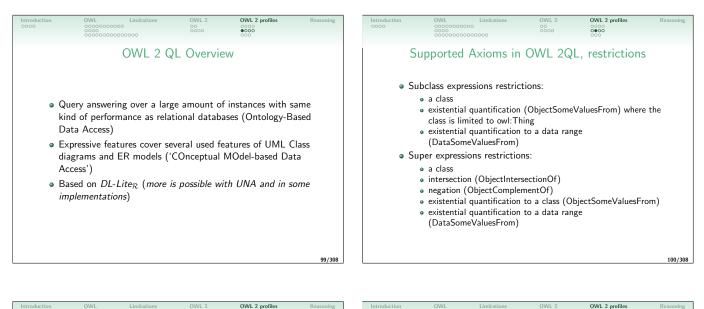




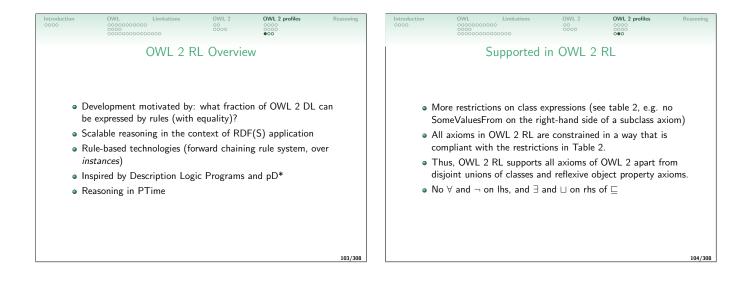
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Partial ta	ble	of f	eatur	es			Rationale	
$\begin{array}{l} Language \Rightarrow \\ Feature \Downarrow \end{array}$	OV Lite	VL 1 DL	OWL 2 DL	OV EL	VL 2 Pro	ofiles RL		
Role hierarchy	+	+	+	1.	+	<u> </u>		
N-ary roles (where $n > 2$)	-	-	-		?			
Role chaining	-	-	+		-			
Role acyclicity	-	-	-		-	· ·		
Symmetry	+	+	+		+		 Computational considerations 	
Role values	-	-	-		-		Computational Considerations	
Qualified number restrictions	-	-	+	-	-		 Consult "OWL profiles" page Table 10. Complexity of 	the
One-of, enumerated classes	?	+	+		-	•		LIIC
Functional dependency	+	+	+	-	?		Profiles	
Covering constraint over concepts	?	+	+		-			
Complement of concepts	?	+	+	-	+		Robustness of implementations w.r.t. scalable applications	ons
Complement of roles	-	-	+		+		• Robustiess of implementations witter semable application	0115
Concept identification	-	-	-		-		Already enjoy 'substantial' user base	
Range typing	-	+	+		+		• Already enjoy substantial user base	
Reflexivity	-	-	+		-			
Antisymmetry	-	-	-		-			
Transitivity	+	+	+		-			
Asymmetry	?	?	+	-	+	+		
Irreflexivity	-	-	+		-	· ·		
· ·	1 .	· ·	· ·		1 · · ·	·		

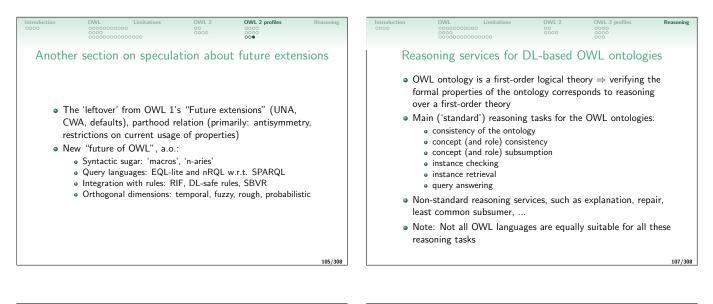


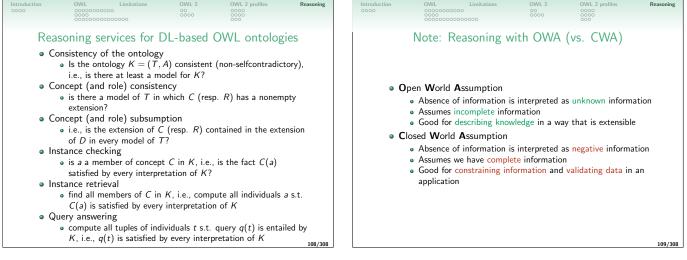


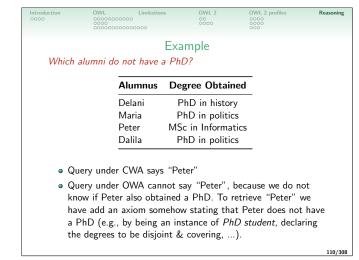


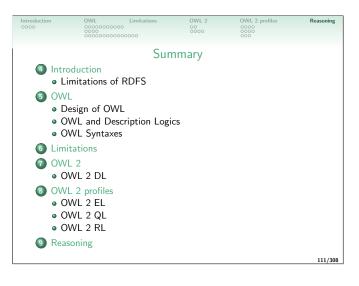


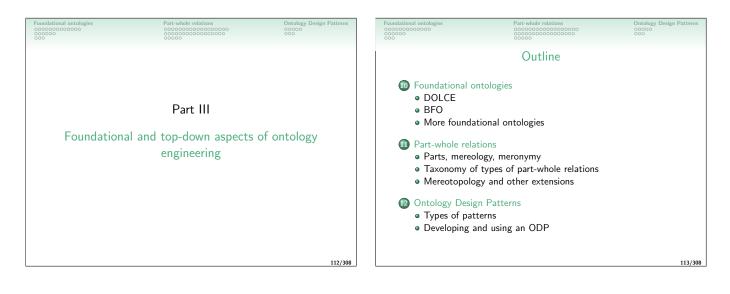


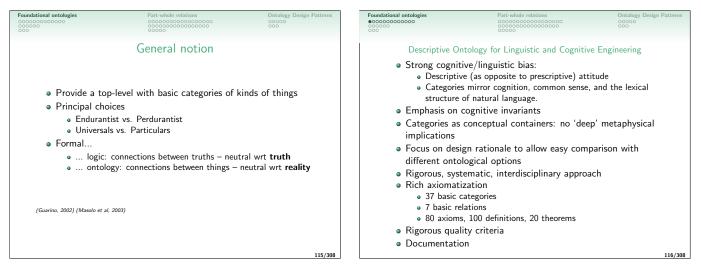


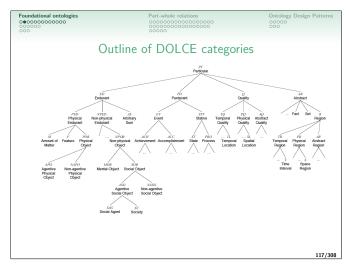


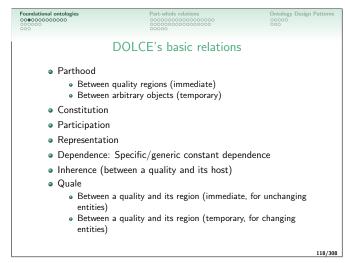


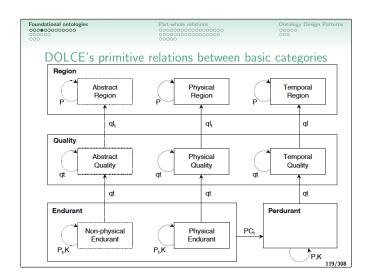


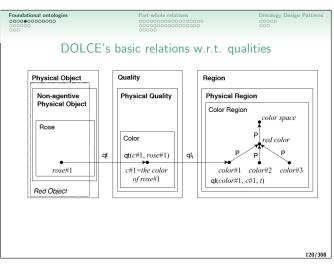


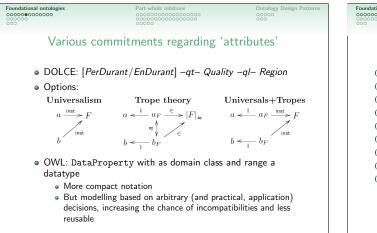


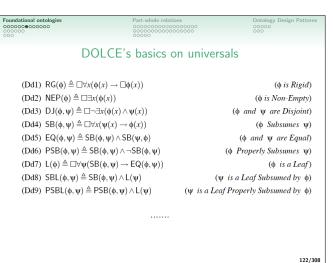


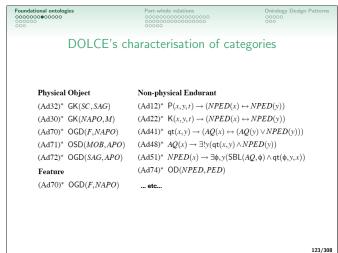


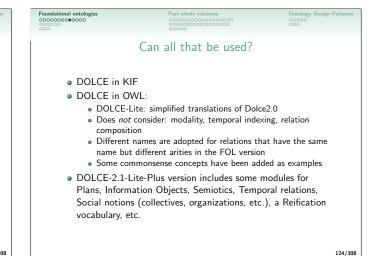






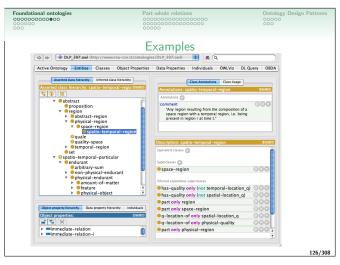


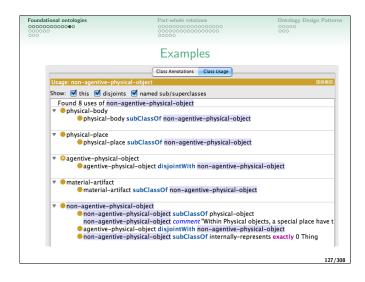


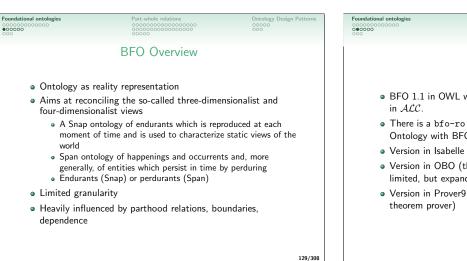


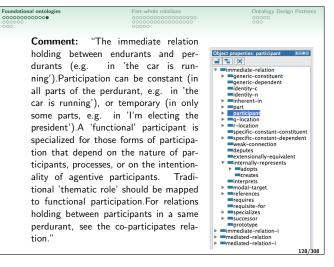
Foundational ontologies	Part-whole relations 000000000000000000 00000000000000000	Ontology Design Patterns 00000 000
	DLP3971	
	es for (re)use: DOLCE-Lite, S ns, ExtendedDnS, and others	ocialUnits,
OWL-DL): Fu properties, etc	nplex to understand (aside fro I DOLCE-Lite-Plus with 208 of (check the "Active ontology" tab in I 7 classes, 70 object properties	classes, 313 object Protégé) and basic
 Time for a DC OWL 2 QL or 	DLCE-Lite ultra- "ultralight"? OWL 2 EL	e.g. for use with
comments classes, an DOLCE, D	DLCE Ultra Lite—DUL—uses fri for classes and properties, has si d includes into a unique file the &S and other modules of DOLC still in OWL-DL (OWL-Lite+Dis	mple restrictions for main parts of Œ Lite+

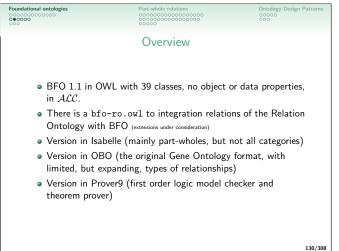


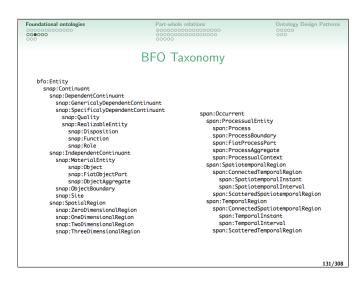


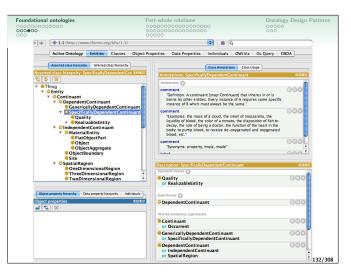




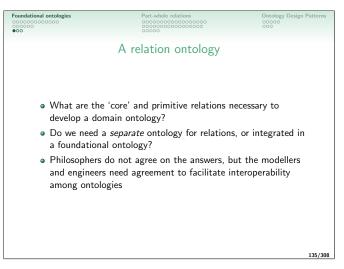


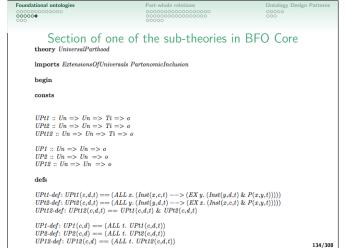




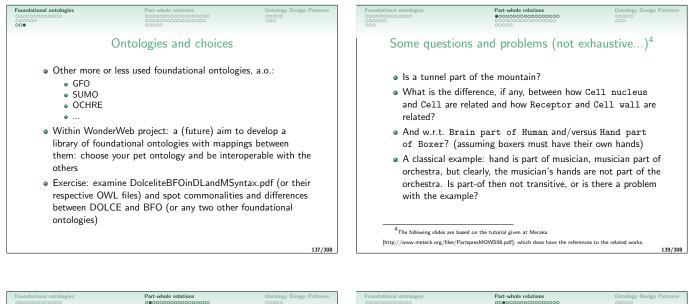


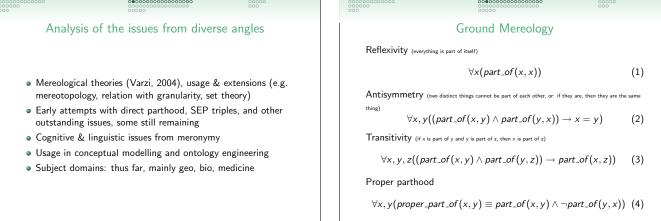
oundational ontologies	Part-whole relations 000000000000000000000000000000000000	Ontology Design Patterns 00000 000
	BFO Core	
and universals BFO as a coll EMR, QS SumsPart	ection of smaller theories izeR, RBG, QDiaSizeR,, Adjac itions, Universals, Instantiation, sOfUniversals, PartonomicInclusic	ency, Collections,
eniversan		





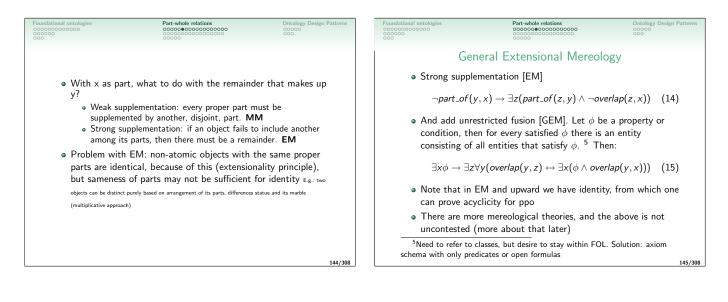


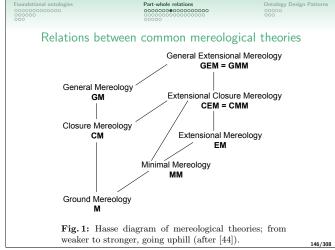




000000000000000000000000000000000000000	00000	sign Patterns
Ground Mereology		
$of(x,y) \equiv part_of(x,y) \land \neg part_of(x,y)$	of(y,x))	(5)
f y then y is not part of x)		
$(part_of(x,y) \rightarrow \neg part_of(y,x))$		(6)
of itself)		
$\forall x \neg (part_of(x, x))$		(7)
	$of(x, y) \equiv part_of(x, y) \land \neg part_of(y, y)$ of y then y is not part of x) $(part_of(x, y) \rightarrow \neg part_of(y, x))$ of itself)	$of(x, y) \equiv part_of(x, y) \land \neg part_of(y, x))$ $f y \text{ then } y \text{ is not part } of x)$ $(part_of(x, y) \rightarrow \neg part_of(y, x))$ of itself)

	000000000000000000000000000000000000000	000 0
Definin	g other relations with <i>part_of</i>	
Overlap (x and y share a p	iece z)	
$\forall x, y(overlap(x))$	$(x,y) \equiv \exists z (part_of(z,x) \land part_of(z,y))$))) (8)
Underlap (x and y are bot	th part of some z)	
$\forall x, y (underlap($	$(x,y) \equiv \exists z (part_of(x,z) \land part_of(y,z))$	·))) (9)
Over- & undercross	6 (over/underlap but not part of)	
$\forall x, y (overcross$	$(x,y) \equiv overlap(x,y) \land \neg part_of(x,y)$) (10)
$\forall x, y (undercross$	$(x,y) \equiv underlap(x,y) \land \neg part_of(y,y)$	x)) (11)
Proper overlap & F	Proper underlap	
$\forall x, y(p_overlap($	$(x, y) \equiv overcross(x, y) \land overcross(y, x)$	<)) (12)
$\forall x, y(p_underlap)$	$p(x,y) \equiv undercross(x,y) \land undercross(x,y)$	(y, x)
		(13)



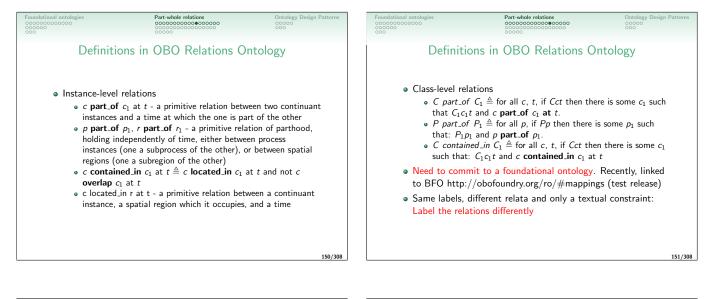


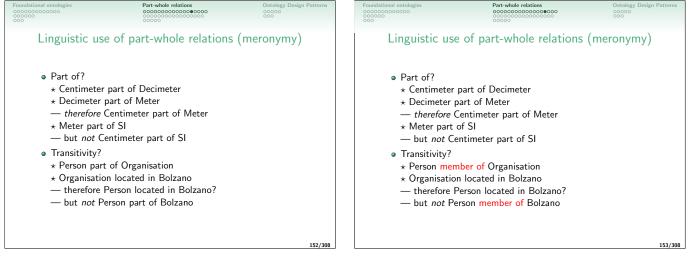
Can any of this be represented in a decidable fragment of first	
order logic for use in ontologies and (scalable) software implementations?	
	7/308

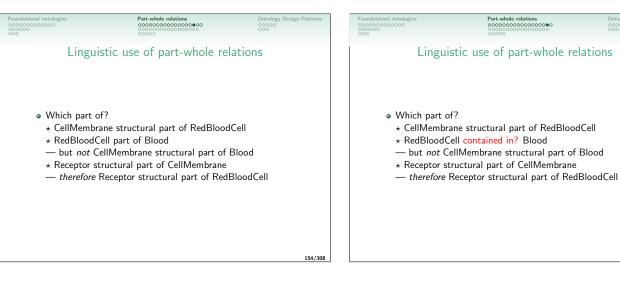
Part-w

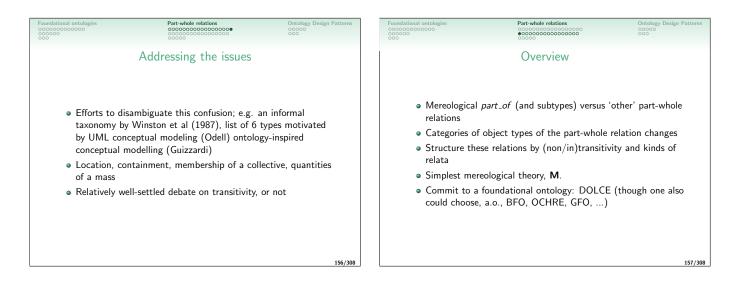
Foundational ontologies	Part-whole relations 000000000000000000000000000000000000	Ontology Design Patterns 00000 000
	Things are improving	
or has-par transitive cl	90s) and simplest options: DL-role t added as primitive role as ≻, mo osure of a parthood relation (16) a ng wheels that in turn have tires (odel it as the and define e.g.
	$\succeq \doteq (\texttt{primitive-part}) *$	(16)
	$\texttt{Car} \doteq \exists \succeq .(\texttt{Wheel} \sqcap \exists \succeq .\texttt{Tire})$	e) (17)
Then Car 🗆	∃≽.Tire	
 SEP triples 	with \mathcal{ALC}	
$ullet$ What \mathcal{SHI}	\mathcal{Q} fixes cf. \mathcal{ALC} : Transitive roles,	Inverse roles (to
have both p	art-of and has-part), Role hierarch	nies (e.g. for
	part-of), qualified Number restrict at a bycicle has-part 2 wheels)	tions (e.g. to
 Build-your-optimized 	wn DL-language	
		148/308

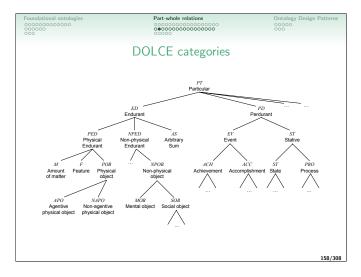
Foundational ontologies	G	art-whole relations	0 000	ology Design Patterns		
What we can(not) implei	ment now v	vith DL-based	d ontology		
X		anguages				
Table: Properties of parthood and proper parthood compared to their support in \mathcal{DLR}_{μ} , \mathcal{SHOIN} and \mathcal{SROIQ} . *: properties of the parthood relation (in M); [‡] : properties of the proper parthood relation (in M).						
			,	,		
Language ⇒	$\overline{\mathcal{DLR}_{\mu}}$	SHOIN	SROIQ	DL-Lite _A		
$\begin{array}{c} \textbf{Language} \Rightarrow \\ \textbf{Feature} \Downarrow \end{array}$,	DL-Lite _A		
Language ⇒		SHOIN	SROIQ	DL-Lite _A		
$\begin{array}{c} \textbf{Language} \Rightarrow \\ \textbf{Feature} \Downarrow \end{array}$		SHOIN	SROIQ	DL-Lite _A		
Language ⇒ Feature ↓ Reflexivity *		SHOIN	SROIQ	,		
Language ⇒ Feature ↓ Reflexivity * Antisymmetry *		SHOIN	SROIQ	DL-Lite _A		
Language ⇒ Feature ↓ Reflexivity * Antisymmetry * Transitivity * ‡		SHOIN	SROIQ	DL-Lite _A		

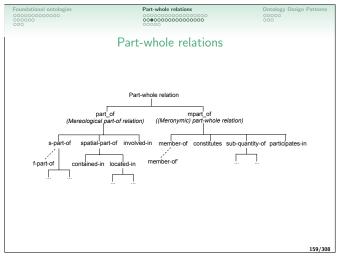


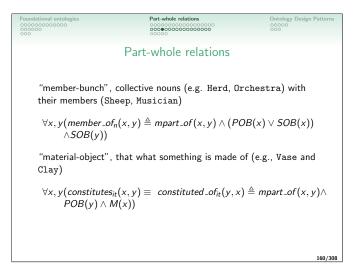


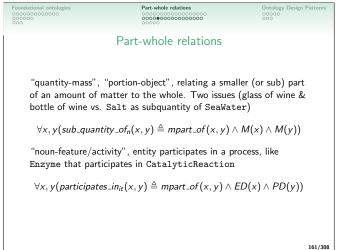


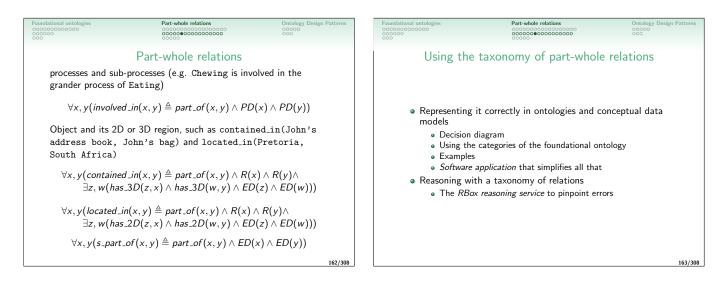


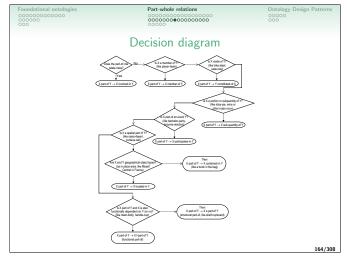


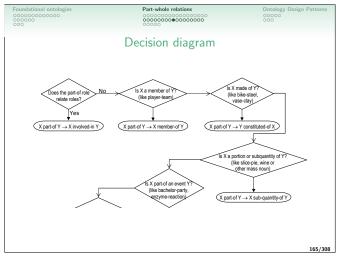


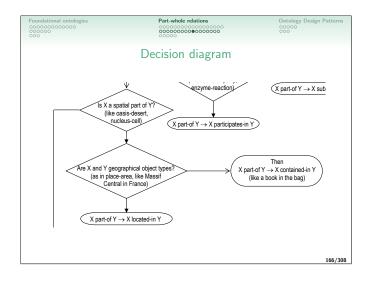


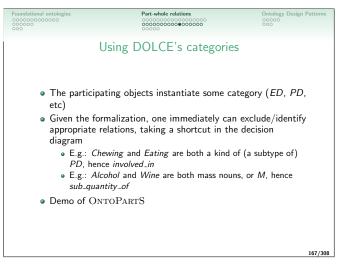


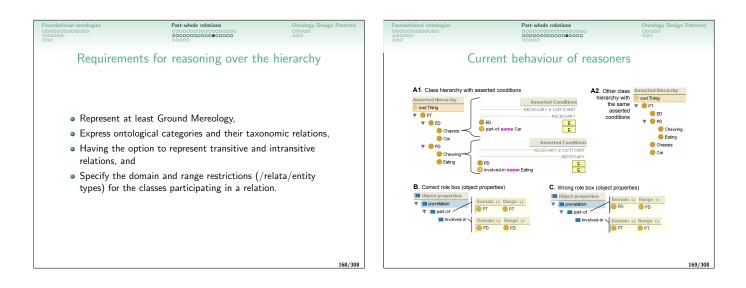


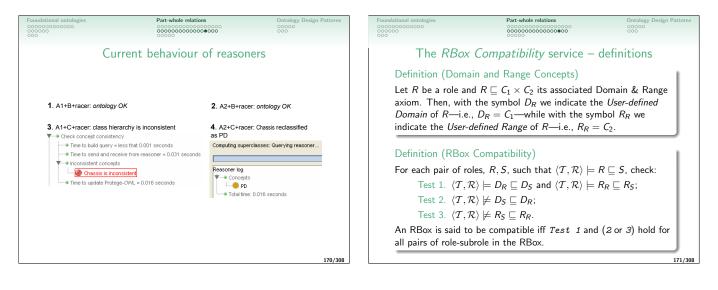


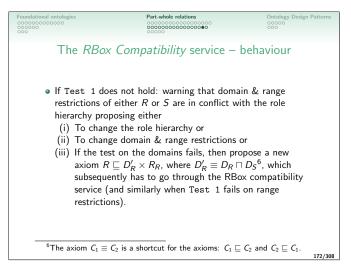


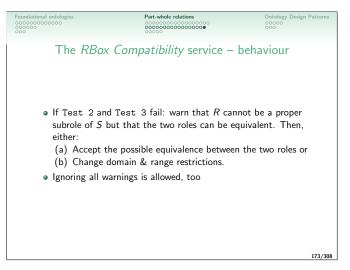


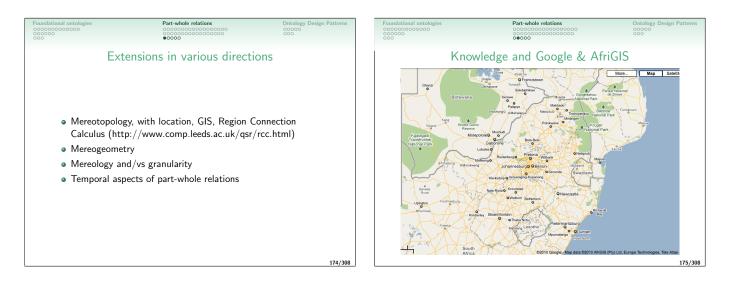


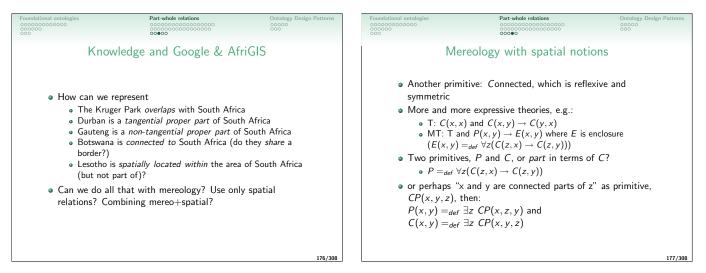


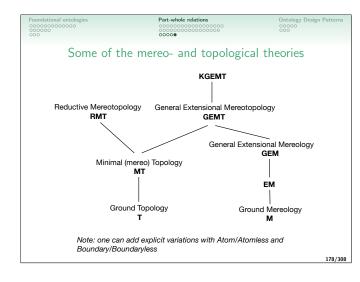


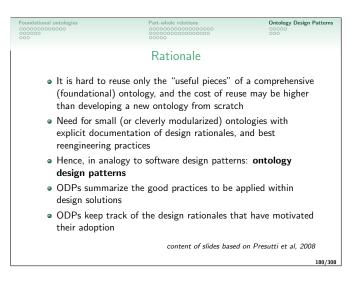






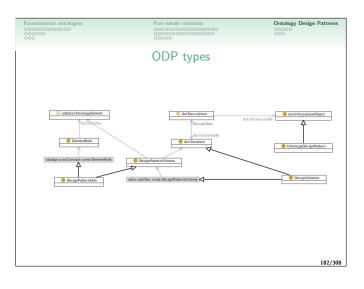






Ontology Design Patterns Ontology Design Patterns Ontology Design Patterns ODDP definition ODP definition ODP definition A design pattern schema is the description of an ODP, including the roles, tasks, and parameters needed in order to solve an ontology design issue An ODP is a modeling solution to solve a recurrent ontology design pattern Schema (or skin) that can only be satisfied by Design Solutions. Design solutions provide the setting for Ontology Elements that play some ElementRole(s) from the

schema. (Presutti et al, 2008)

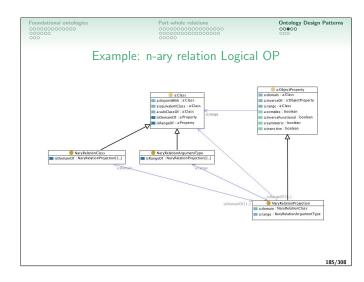


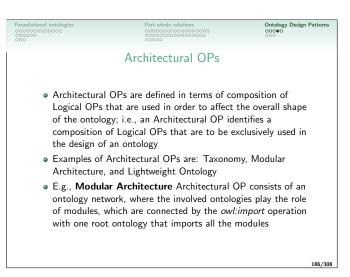
Types of Patterns Structural OPs • Six families of ODPs: Structural OPs, Correspondence OPs, Content OPs (CPs), Reasoning OPs, Presentation OPs, and Lexico-Syntactic OPs Logical OPs: • Are compositions of logical constructs that solve a problem of • CPs can be distinguished in terms of the domain they expressivity in OWL-DL (and, in cases, also in OWL 2 DL) represent Only expressed in terms of a logical vocabulary, because their • Correspondence OPs (for reengineering and mappings-next signature (the set of predicate names, e.g. the set of classes lecture) and properties in an OWL ontology) is empty Independent from a specific domain of interest Reasoning OPs are typical reasoning procedures • Logical macros compose OWL DL constructs; e.g. the universal+existential OWL macro • Presentation OPs relate to ontology usability from a user perspective; e.g., we distinguish between Naming OPs and • Transformation patterns translate a logical expression from a Annotation OPs logical language into another; e.g. n-aries • Lexico-Syntactic OP are linguistic structures or schemas that

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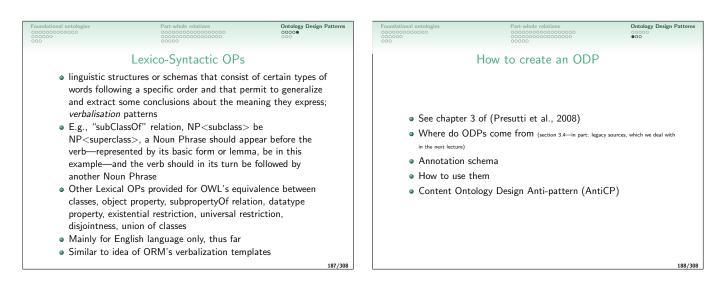
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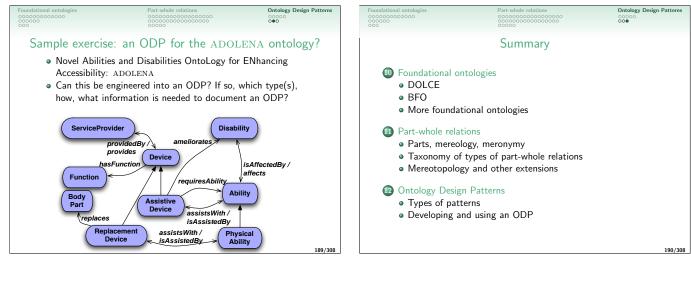
permit to generalize and extract some conclusions about the meaning they express

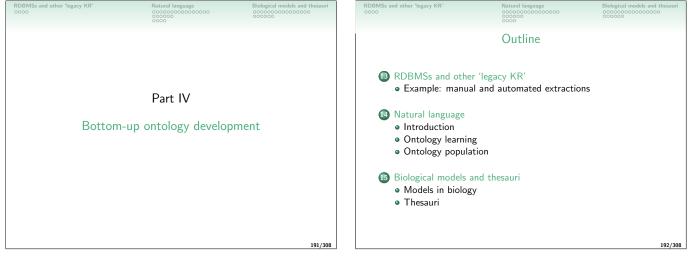




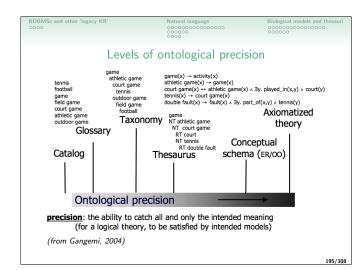
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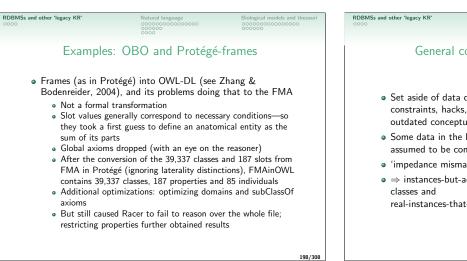


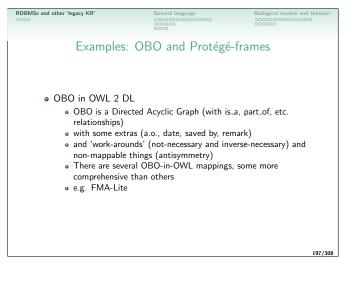


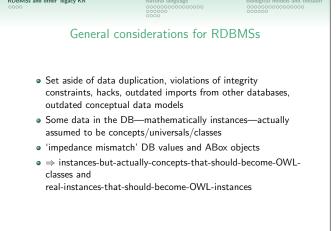






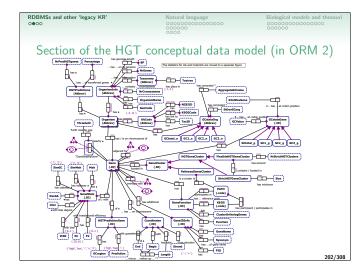


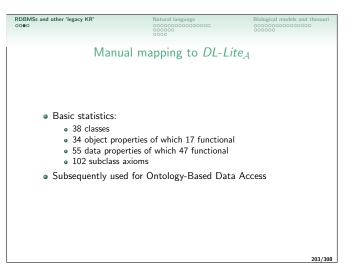


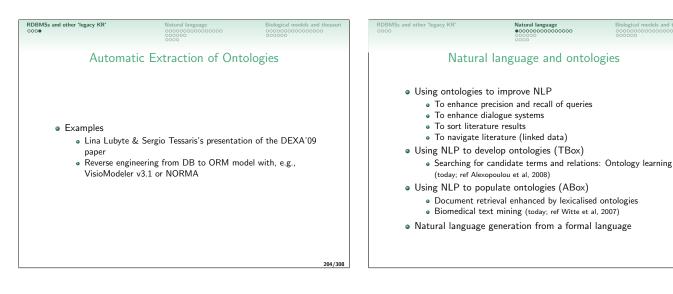


RDBMSs and other 'legacy KR RDBMSs and other 'legacy KR General considerations for RDBMSs Manual Extraction • Reuse/reverse engineer the physical DB schema • Reuse conceptual data model (in ER, EER, UML, ORM, ...) • But, • Most database are not neat as assumed in the 'Automatic · Assumes there was a fully normalised conceptual data model, Extraction of Ontologies' (e.g., denormalised) Denormalization steps to flatten the database structure, which. • Then what? if simply reverse engineered, ends up in the ontology as a class with umpteen attributes • Reverse engineer the database to a conceptual data model Minimal (if at all) automated reasoning with it · Choose an ontology language for your purpose • Redo the normalization steps to try to get some structure • Example: the HGT-DB about horizontal gene transfer (the back into the conceptual view of the data? same holds for the database behind ADOLENA) • Add a section of another ontology to brighten up the 'ontology' into an ontology? • Establish some mechanism to keep a 'link' between the terms in the ontology and the source in the database?

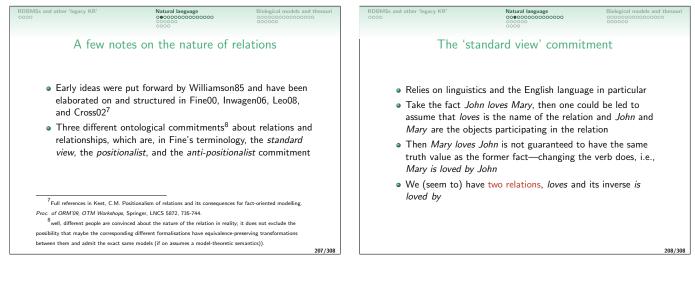
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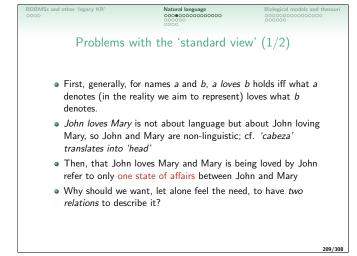


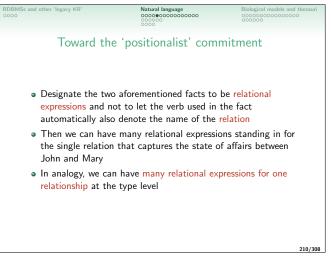


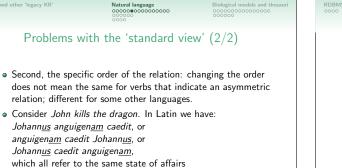


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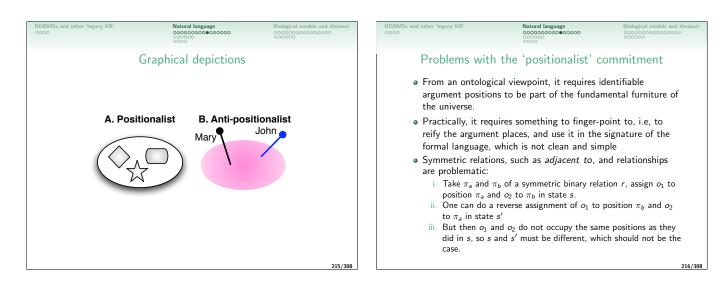


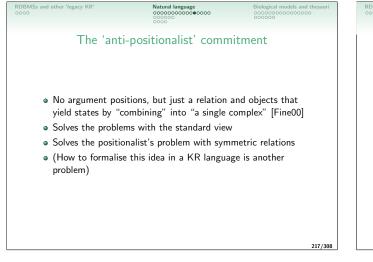
- But Johannum anguigena caedit is a different story alltogether
- Likewise for John loves Mary and Johannus Mariam amat versus Johannum Maria amat.

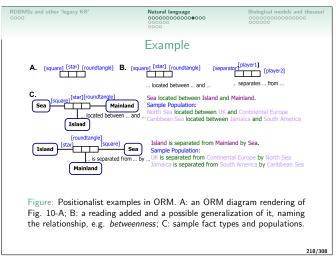
A linguistic version of argument places (roles) thanks to the nominative and the accusative that are linguistically clearly indicated The order of the argument places is not relevant for the relation itself English without such declensions that change the terms so as to disambiguate the meaning of a relational expression Inverses for seemingly asymmetrical relations necessarily exist in reality and descriptions of reality in English, but not in other languages even when they represent the same state of affairs??? Asymmetric relational expressions, but this does not imply

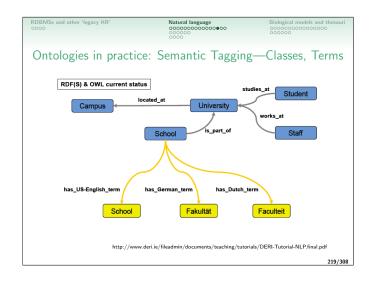
that the relation it verbalises is asymmetric

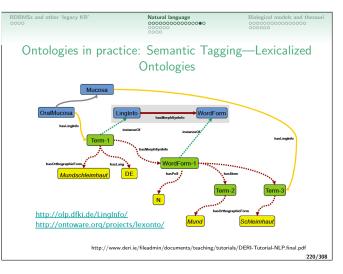
Natural lang Natural lan The 'positionalist' commitment The 'positionalist' commitment Ingredients (i) an *n*-ary relationship R with A_1, \ldots, A_m participating object • Binary relation killing and identify the argument types (m < n), places—"argument positions" [Fine00] to have (ii) *n* argument places π_1, \ldots, π_n , and "distinguishability of the slots" [Cross02]-killer and deceased (iii) n assignments α_1,\ldots,α_n that link each object $\textit{o}_1,\ldots,\textit{o}_n$ (each (loosely, a place for the nominative and a place for the object instantiating an A_i) to an argument place ($\alpha \mapsto \pi \times o$) accusative), assign John to killer and the dragon to deceased • R, $\pi_1,$ $\pi_2,$ $\pi_3,$ $r\in R,$ $o_1\in A_1,$ $o_2\in A_2,$ $o_3\in A_3,$ then any of and order the three elements in any arrangement $\forall x, y, z(R(x, y, z) \rightarrow A_1(x) \land A_2(y) \land A_3(z))$ and its • Relation(ship) and several distinguishable 'holes' and we put permutations with corresponding argument places-i.e., each object in its suitable hole. $R[\pi_1, \pi_2, \pi_3]$, and e.g., $R[\pi_2, \pi_1, \pi_3]$, and $[\pi_2\pi_3]R[\pi_1]$ —all • There are no asymmetrical relations, because a relationship Rdenote the same SoA under the same assignment \textit{o}_1 to π_1, \textit{o}_2 to π_2 , and o_3 to π_3 for the extension and its inverse R^- , or their instances, say, r and r', are identical, i.e., the same thing [Williamson85,Fine00,Cross02] • Thus, $r(o_1, o_2, o_3)$, $r(o_2, o_1, o_3)$, and $o_2 o_3 r o_1$ are different representations of the same SoA where objects o_1 , o_2 , and o_3 are related to each other by means of relation r. 213/308 214/308











Ss and other 'legacy KR'	Natural language 000000000000000 00000 00000000000000 00000 000000000000000 00000	Biological models and thesauri	RDBMSs and other 'legacy KR' 0000	Natural language ○○○○○○○○○○○○○○○○ ●○○○○○ ○○○○	Biological models and coccococococococo coccoco
Exa	amples (out of many)		Background	
	<pre>e http://www.deri.ie/fileadmi /DERI-Tutorial-NLP.final.pdf t al, 2009)</pre>		 Ontology develop 	oment is time consuming	
bio(medical)	bMed, which indexes \pm 19m domain; pre-processing of the nantic tagging)		 Bottom-up ontol to use NLP 	ogy development strategie	es, of which one is
· · · · · · · · · · · · · · · · · · ·	PubMed query are sorted ac	cording to terms in	 Where, if anywhere, if anywhere, if anywhere, and development, and 	ere, can NLP make life ea d how?	sier for ontology
 Question answer 2009) 	system AliQAn for agricult	UIP (Vila and Ferrández,		re mostly discouraging, an que, and ontological comr	
	gnment task too difficult for to an open domain QA syste nd WordNet			oser look at ontology learnin omain ontology	g limited to finding
 Attempto Contro engine, template- 	lled English (ACE), rabbit, based approach	etc.; grammar			
		221/308			

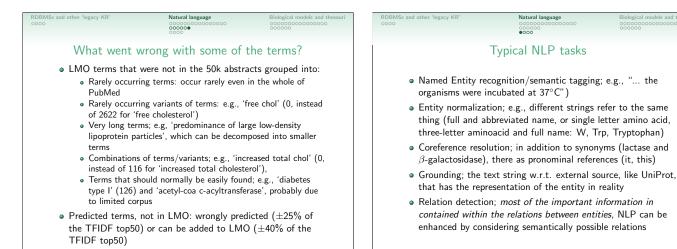
Bottom-up ontology development with NLP

- Usual parameters, such as purpose (in casu, document retrieval), formal language (an OWL species)
- A standard kind of ontology (not a comprehensive lexicalised ontology)
- Additional considerations for "text-mining ontologies"
 - Level of granularity of the terms to include (hypo/hypernyms)
 - How to deal with synonyms ('LDL I' and 'large LDL')
 Handle term variations (e.g., 'LDL-I' and 'LDL I', 'Tangiers'
 - Handle term variations (e.g., LDL-1 and LDL 1, Tang disease' and 'Tangier's Disease')
 - Disambiguation; e.g. w.r.t. abbreviations

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222/308 Natural language 000000 Method to test automated term recognition • Compare the terms of a manually constructed ontology with $\stackrel{\cdot}{\text{terms}}$ obtained from text mining a suitable corpus Build an ontology manually Lipoprotein metabolism (LMO), 223 classes with 623 synonyms • Create a corpus 3066 review article abstract from PubMed, obtained with a 'lipoprotein metabolism' search • Automatic Term Recognition (ATR) tools • Text2Onto: relative term frequency, TFIDF, entropy, hypernym structure of WordNet, Hearst patterns · Termine: statistics of candidate term, such as total frequency of occurrence, frequency of the term as part of other longer candidate terms, length of term OntoLearn: linguistic processor and syntactic parser, Domain relevance and domain consensus • RelFreq: relative frequency of a term in a corpus • TFIDF: RelFreq + doc. frequency derived from all phrases in PubMed 224/308

DBMSs and other 'legacy KR' 200	Natural language 000000000000000 000000 0000	Biological models and thesauri	RDBMSs and other 'legacy KR' 0000	Natural languag		Biological models and thesau 000000000000000000000000000000000000
	Results		R	esults (cor	nt'd)	
terms	led form analysis because ncluded in analysis for up	C .	Table 3: Coverage of LMO terminology in selecte mining: Even a large text base with 50,000 docum	ents contains only 71% o		
	5 17-35% for top 50 term	s and 4-8% for top	·	1000	all	contained
1000 terms Precision for LMO	D + expert analysis of the up to 75% for top 50 ter	automatically	300 review abstracts for "lipoprotein metabolism" 3.064 abstracts for "lipoprotein metabolism" 50,000 abstracts containing "lipoprotein"	8.75% 14.99%	15.35% 38.25%	20.98% 53.00% 71.22%
	the longer terms, RelFree	and TFIDF for			fr	om Alexopoulou et al. 2008
 Termine good for the shorter terms 						





Requirements for NLP ontologies

- Domain ontology (at least a taxonomy)
- Text model, concerns with classes such as sentence, text position and locations like abstract, intorduction
- · Biological entities, i.e., contents for the ABox, often already available in biological databases on the Internet
- Lexical information for recognizing named entities; full names of entities, their synonyms, common variants and misspellings, and knowledge about naming, like endo- and -ase
- Database links to connect the lexical term to the entity represent in a particular database (the grounding step)
- Entity relations; represented in the domain ontology

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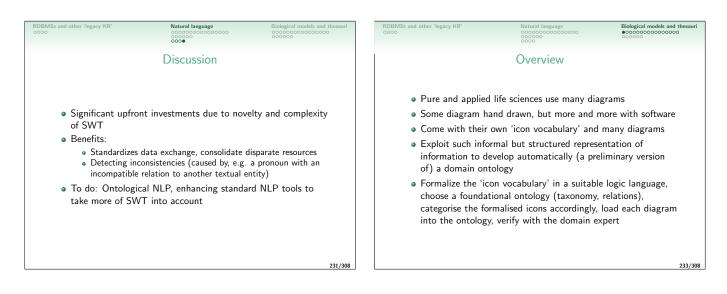
• See Witte et al. book chapter for details • Ontology in OWL, in Protégé; with class name, textual definition and example instances • Species info from the NCBI taxonomy; note the management of central scientific name and its synonyms, common variants • Uniprot and use of its back-links to the NCBI taxonomy

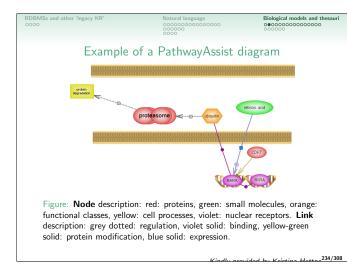
Natural languag

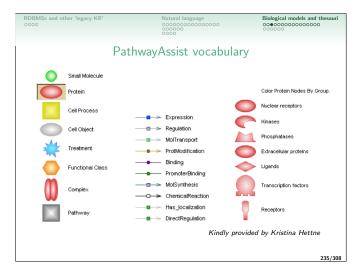
MutationMiner use case

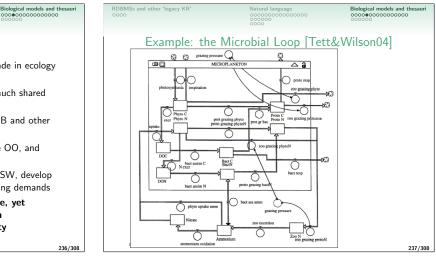
and misspellings

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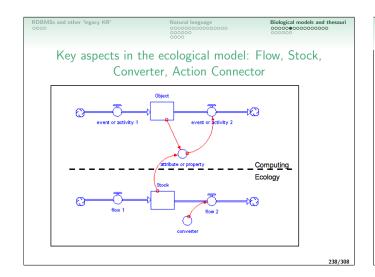


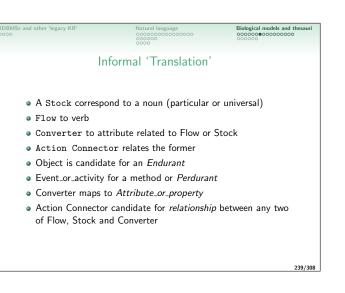


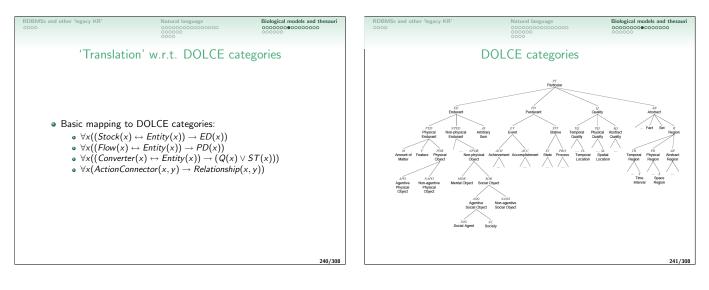


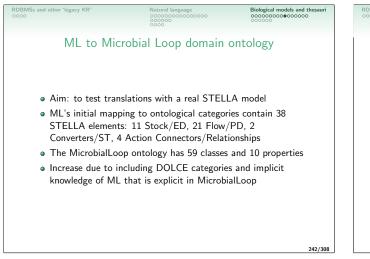
Case study motivation

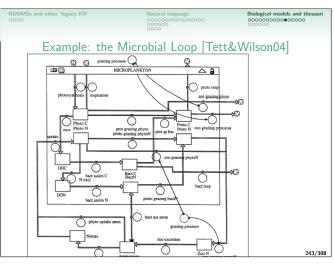
- Experiment in 2005 (Keet, 2005), but progress made in ecology (Madin et al, 2008; MTSR'09 proceedings)
- Extensive use of modelling in ecology, but not much shared (depending on sub-discipline)
- Models used with independent software tools (DB and other applications)
- 'Legacy code' (procedural), moving toward more OO, and ontologies
- Requirement for (re-)analysis to upgrade legacy SW, develop new SW to meet increasing complexities and rising demands
- use the opportunity to create a more durable, yet computationally usable, shared, agreed upon representation of the knowledge about reality



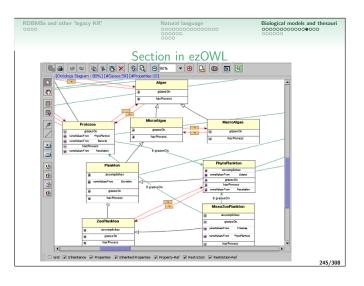


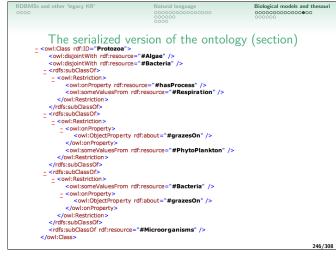


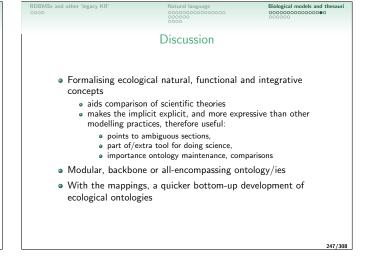


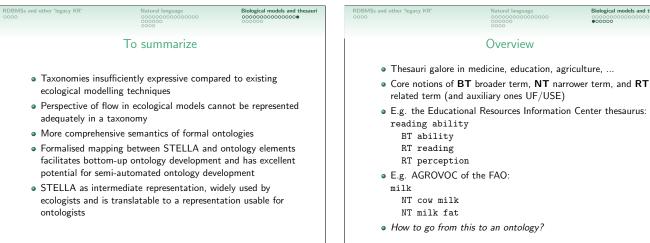


)	R'	Natural language 000000000000000 00000 0000	Biological models and the	
Section of n	nore re	fined mapping to DOC	CLE categories	
Phyto C	NAPO	Phyto C = phytoplankton organic carbon. Phytoplankton is an APO, but 'phyto C' is <i>part</i> of the APO: only the organic carbon of the phytoplankton, not the organism as an active agent as such		
Phyto N	NAPO	Phyto N = phytoplankton nitrogen		
DOC	NAPO	DOC = detrital organic carbon. Detritus is an E amount of matter (M), but here, like with the or on only a <i>part</i> of the NAPO		
Nitrate	NAPO	Dissolved nitrate. Molecules are non agentive p	hysical objects.	
Flow				
Photosynthesis	PRO	To phytoplankton N		
Respiration	PRO	From phytoplankton N		
Prot gr bac	PRO	Protozoa that are grazing on the Bacter	rial C	
Converter				
Grazing pressure	ST	Acts on a PRO affecting the process of pressure' is there (might reach zero), h		
Action connector				
117	Yes	Acts on the mesozooplankton grazing on the pr mesozooplankton grazing on the pr hasGrazingPressure	rotozoa, and acts on the hytoplankton: relation	





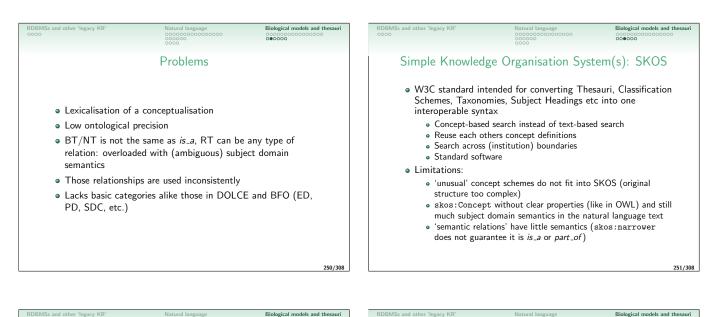


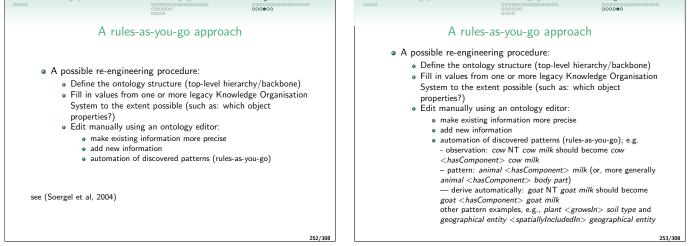


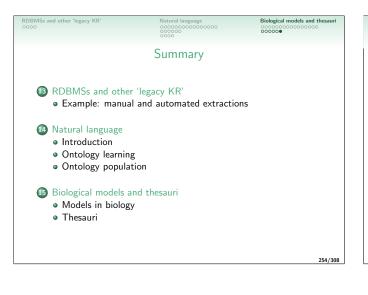
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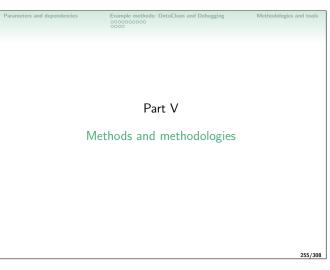
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Biological models and thesauri











The landscape

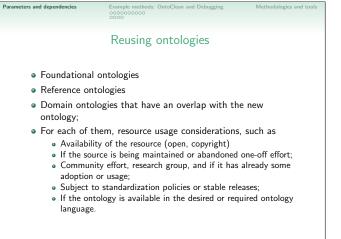
- Multiple modelling issues in ontology development for the applied life sciences (e.g., part-of, uncertainty, prototypes, multilingual), methodological issues, highly specialised knowledge
- W3C's incubator group on modelling uncertainty, mushrooming of bio-ontologies, ontology design patterns, W3C standard OWL, etc.
- Solving the early-adopter issues moves the goal-posts
 - Which ontologies are reusable for one's own ontology?
 - What are the consequences choosing one ontology over the other?
 - The successor of OWL, draft OWL 2, has 5 languages: which one should be used for what and when?

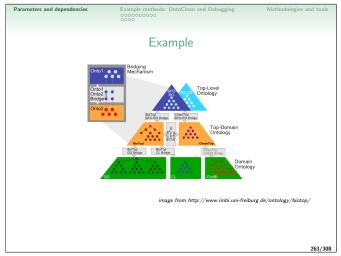
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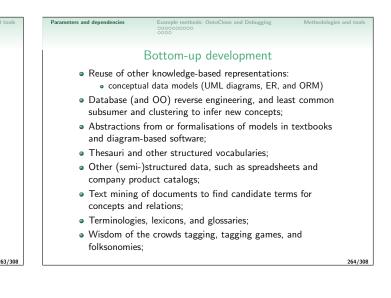
Parameters and dependencies Example methods: OntoClean and Debugging Methodologies and tools Ontology in an ontology-driven information system destined for run-time usage, e.g., in scientific workflows, MASs, ontology-mediated data clustering, and user interaction in e-learning Ontologies for NLP, e.g.m annotating and querying Digital Libraries and scientific literature, QA systems, and materials for e-learning • As full-fledged discipline "Ontology (Science)", where an ontology is a formal, logic-based, representation of a scientific theory • Tutorial ontologies, e.g., the wine and pizza ontologies

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Purposes Querying data by means of an ontology (OBDA) through linking databases to an ontology Database integration, (GO, OBO Foundry) Structured controlled vocabulary to link data(base) records and navigate across databases on the Internet ('linked data') Using it as part of scientific discourse and advancing research at a faster pace, (including experimental ontologies) Coordination among and integration of Web Services







Languages – preliminary considerations Depending on the purpose(s) (and available resources), one ends up with either (a) a large but simple ontology, i.e., mostly just a taxonomy without, or very few, properties (relations) linked to the concepts, where 'large' is, roughly, > 10000 concepts, so that a simple representation language suffices; (b) a large and elaborate ontology, which includes rich usage of properties, defined concepts, and, roughly, requiring OWL-DL;

- (c) a small and very complex ontology, where 'small' is, roughly, < 250 concepts, and requiring at least OWL 2 DL
- Certain choices for reusing ontologies or legacy material, or goal, may lock one a language
- $\bullet \; \Rightarrow$ Separate dimension that interferes with the previous parameters: the choice for a representation language

or

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Languages Older KR languages (frames, obo, conceptual graphs, etc.) Web Ontology Languages: OWL: OWL-Lite, OWL-DL, OWL full OWL 2 with 4 languages to tailor the choice of ontology language to fit best with the usage scope in the context of a scalable and multi-purpose SW: OWL 2 DL is most expressive and based on the DL language SROTQ OWL 2 QL fragment to achieve better performance with larger ontologies (e.g., for use with SNOMED-CT) OWL 2 QL fragment to achieve better performance with ontologies linked to large amounts of data in secondary storage (databases); e.g. DIG-QuOnto OWL 2 RL has special features to handle rules Extensions (probabilistic, fuzzy, temporal, etc.)

• Differences between expressiveness of the ontology languages and their trade-offs

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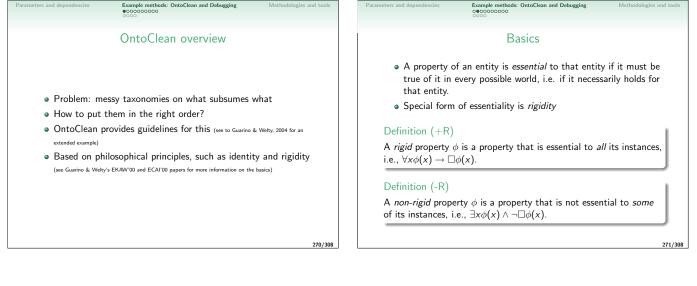
 Perameters and dependencies
 Example methods: OntoClean and Debugging
 Methodologies and tools

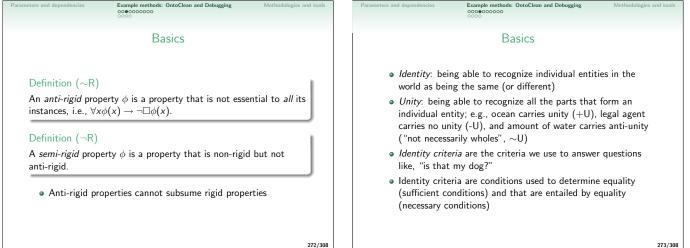
 Bescription
 Reasoning services

 • Description logics-based reasoning services
 • The standard reasoning services for ontology usage: satisfiability and consistency checking, taxonomic classification, instance classification;
 • 'Non-standard' reasoning services to facilitate ontology development: explanation/justification, glass-box reasoning, pin-pointing errors, least-common subsumer;
 • Querying functionalities, such as epistemic and (unions of) conjunctive queries;

 • Ontological reasoning services (OntoClean, RBox reasoning service)
 • Other technologies (e.g., Bayesian networks)

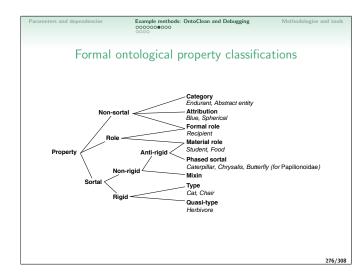


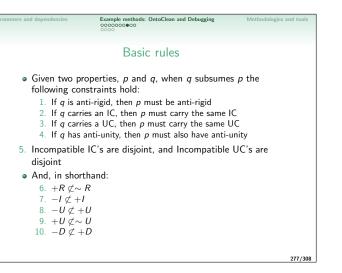


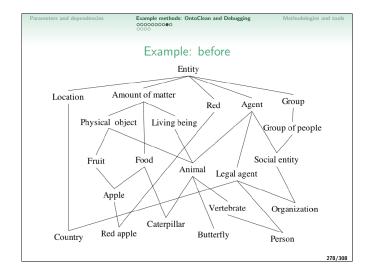


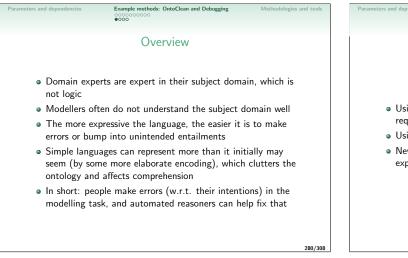
ameters and dependencies	Example methods: OntoClean and Debugging 0000@00000 0000	Methodologies and tools
	Basics	
Definition		1
A non-rigid prope property carrying	rty carries an IC Γ iff it is subsumed Γ .	l by a rigid
Definition		
iii) Γ is not carrie that, if ϕ inherits	blies an IC Γ iff i) it is rigid; ii) it ca d by all the properties subsuming ϕ different (but compatible) ICs from counts as supplying an IC.	. This means
 Any property 	carrying an IC: +I (-I otherwise).	
a Any property	supplying an IC: $+O$ (-O otherwise	e); "O" is a
	r "own identity"	

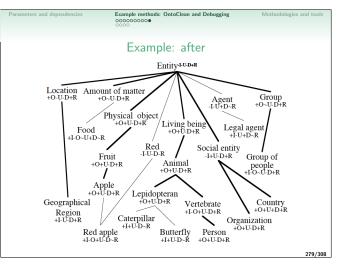
Parameters and dependencies		00	ample me	thods: OntoClean and Debugg	ing Methodo	Methodologies and tools	
	For	mal	onto	logica	al property clas	ssifications	
	+0	+1	+R	+D -D	Туре	Sortal Non-Sortal	
	-0	+1	+R	+D -D	Quasi-Type		
	-0	+1	~R	+D	Material role		
	-0	+	~R	-D	Phased sortal		
	-0	+I	⊣R	+D -D	Mixin		
	-0	-1	+R	+D -D	Category		
	-0	-1	~R	+D	Formal role		
	-0	-1	~R ⊣R	-D +D -D	Attribution		
		•		•		·	275/308

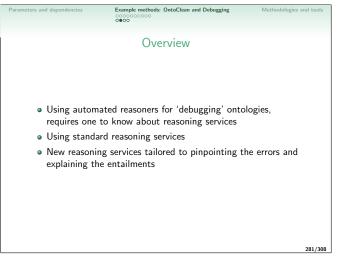








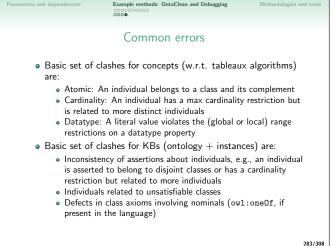




Common errors

- Unsatisfiable classes
 - In the tools: the unsatisfiable classes end up as direct subclass of owl:Nothing
 - Sometimes one little error generates a whole cascade of unsatisfiable classes
- Satisfiability checking can cause rearrangement of the class tree and any inferred relationships to be associated with a class definition: 'desirable' vs. 'undesireable' inferred subsumptions
- Inconsistent ontologies: all classes taken together unsatisfiable

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Example methodology: METHONTOLOGY Where are we? • Basic methodology: • Parameters that affect ontology development, such as · specification: why, what are its intended uses, who are the prospective users purpose, base material, language conceptualization, with intermediate representations • Methods, such as reverse engineering text mining to start, · formalization (transforms the domain-expert understandable OntoClean to improve conceptual model' into a formal or semi-computable model) • Tools to model, to reason, to debug, to integrate, to link to implementation (represent it in an ontology language) maintenance (corrections, updates, etc) data • Additional tasks (as identified by METHONTOLOGY) • Methodologies that are coarse-grained: they do not (yet) · Management activities (schedule, control, and quality contain all the permutations at each step, i.e. what and how assurance) to do each step, given the recent developments; Support activities (knowledge acquisition, integration, • e.g. step x is "knowledge acquisition", but what are it evaluation, documentation, and configuration management) component-steps? • Applied to chemical, legal domain, and others (More comprehensive nent of extant methodologies in Corcho et al, 2003)

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Methodologies and tool

Extending the methodologies

- METHONTOLOGY, MoKi, and others (e.g., On-To-Knowledge, KACTUS approach) are for developing one *single* ontology
- Changing landscape in ontology development towards building "ontology networks"
- Characteristics: dynamics, context, collaborative, distributed
- E.g., the emerging NeOn methodology

- MoKi is based on a SemanticWiki, which is used for collaborative and cooperative ontology development
- It enables actors with different expertise to develop an "enterprise model"⁹: use both *structural (formal) descriptions* and *more informal* and *semi-formal* descriptions of knowledge

MOdelling wiKI

- $\bullet \; \Rightarrow \;$ access to the enterprise model at different levels of formality: informal, semi-formal and formal
- more info and demo at http://moki.fbk.eu

⁹ enterprise model: "a computational representation of the structure, activities, processes, information, ources, people, behavior, goals, and constraints of a business, government, or other enterprise"

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Methodologies and tool

Extending the methodologies: NeOn

- NeOn's "Glossary of Activities" identifies and defines 55 activities when ontology networks are collaboratively built
- Among others: ontology localization, -alignment, -formalization, -diagnosis, -enrichment etc.
- Divided into a matrix with "required" and "if applicable"
- Embedded into a comprehensive methodology (under development)

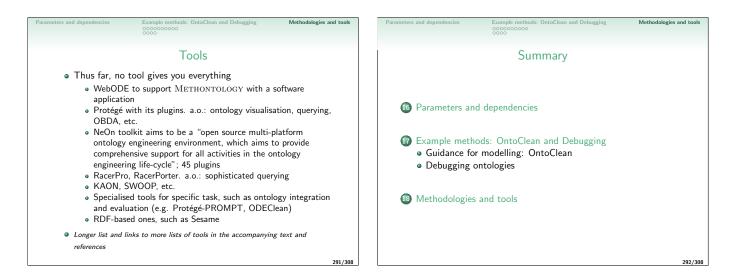
(more info in neon_2008_d5.4.1.pdf)

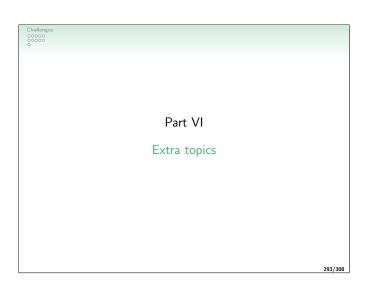
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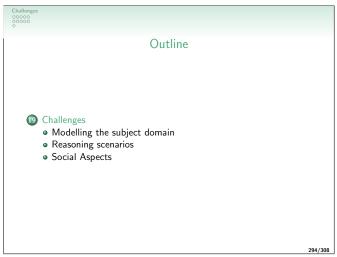
Methodologies and to

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ories and tool







SWT challenges or failures?

- 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?
- A failure according to one may be considered a challenge by another
- Offer and demand, perceptions, perspectives, expectations

A few general issues

- 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

Which language do we need?

• The (reflexive, antisymmetric, transitive) parthood relation

• Each plane passenger boards the aircraft after having checked

• Each Government has as members at least 10 Ministers

• Any two people are related to each other in one way or

A father is necessarily male

Swedish people are very tall

· Generally, birds do fly

• The class of people who are young

• 90% of the Italians have brown eyes

in

another

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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):
 - $\bullet\,$ to some, there is more in OWL/OWL2 than needed and used $\bullet\,$ to some, there is not enough
- 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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Challenge

Limitations as identified by users/modellers (a.o., Schulz et al, 2009)

- *n*-ary relations, where *n* > 2
- "Hepatitis hasSymptom Fever in most but not all cases"
 - What about doing it with probabilistic default knowledge? • $(\psi \mid \phi)[I, u]$ as "generally, if an object belongs to ϕ , then it
 - belongs to ψ with a probability in [I, u]"
 - e.g., $(\exists hasSymptom.Fever | Hepatitis)[1,1]$
- "In 2000, worldwide prevalence of diabetes mellitus was 2.8%"
 - Probabilistic, or arithmetic, or what have we?
 - First, it assumes some class Human and a class
 - HumanDiabetesMellitus, where some of the instances of the former have (are bearerOf) an instance of the latter
 - Second, we have some notion of prevalence, but what is it associated to (a property of)? of the human *population* in the world, not a property of an individual human

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Challenges 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. "[Diabete:Human] = 0.028" Another option: put in TBox with a data property, e.g., HumanDiabetesMellitus 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?

- Problems with Drug Abuse Prevention (in SNOMED CT)
 - DrugAbusePrevention □ Procedure □ ∃hasFocus.DrugAbuse
 DrugAbusePrevention ≡ Procedure □ ∃hasParticipant.Person □ ∃causes.(State □ hasParticipant.(Person □ ∃participatesIn.¬ DrugAbuse))

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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
 - "aspirin prevents myocardial infarction"
 - $\bullet\,$ 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 □⁺
 - e.g., AspirinIntake ⊆ □⁺prevents.MyocardialInfarction, Or MyocardialInfarction ⊑ □⁺preventedBy.AspirinIntake, Or AspirinIntake ⊑
 - $\square^+ \texttt{hasPhysiologicalEffect.} \neg \texttt{MyocardialInfarction} ~?$

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• Usual model checking

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Introduction on reasoning scenarios

- ${\scriptstyle \bullet}\,$ The standard reasoning services are obviously sorted out
- Performance issues for the 'debugging' and explanation reasoning, and how to provide the 'best' explanation
- $\bullet\,$ Querying OWL 2 DL, and any ABox data
- Additional reasoning scenarios with 'standard' ontologies
- Reasoning over fuzzy, rough, probabilistic, possibilistic, time, ontologies and data

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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
- 9. Complex queries

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

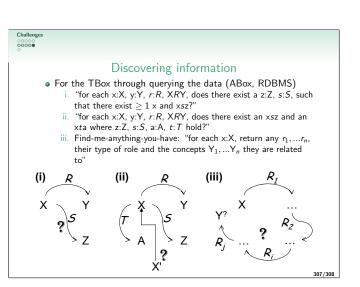
Checking against instances

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
 - else in the ontology



Building ontologies involves humans

- Building an ontology is, generally, an *interdisciplinary* (transdisciplinary?) endeavour
- Different disciplines with different mores, goals
- The collaboration requires patience, respect, capability to listen, compromise
- More slides in a separate file, time permitting

Challenges