

Approximation for the Semantic Web

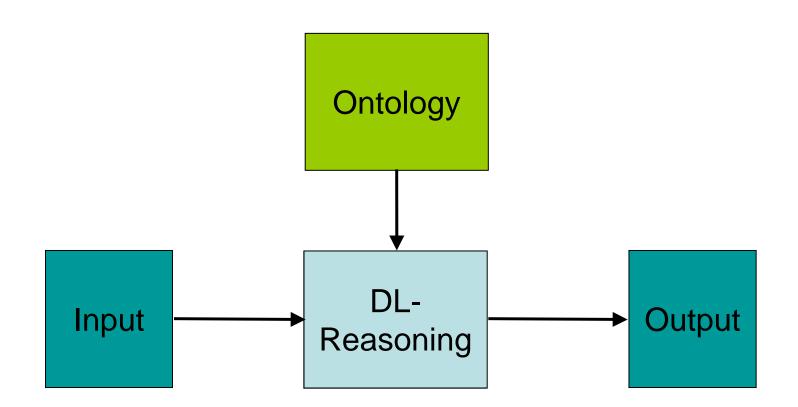
The KnowledgeWeb point of view

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KMI Podium, Milton Keynes, May 5th 2006



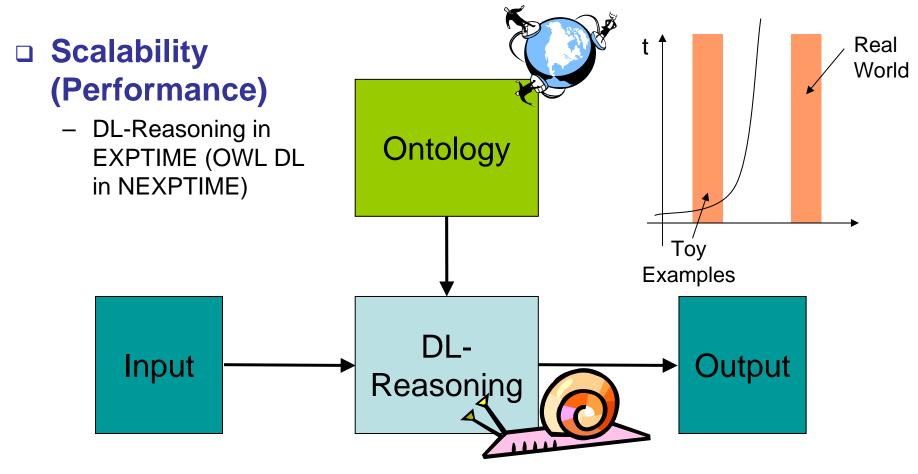








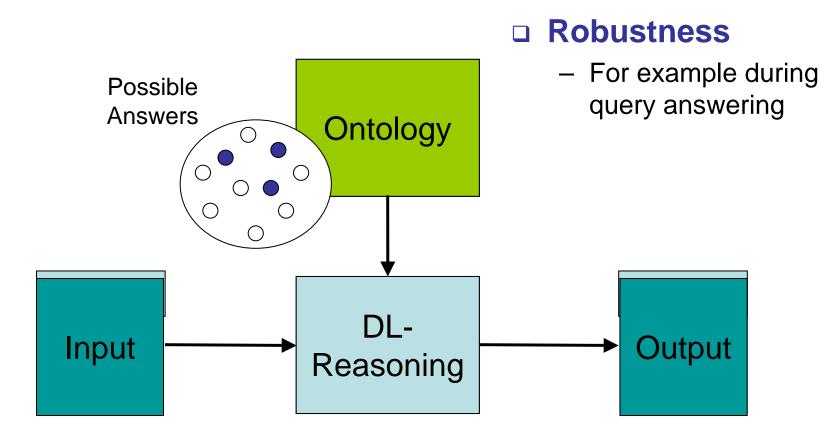
Problems tackled in KWEB





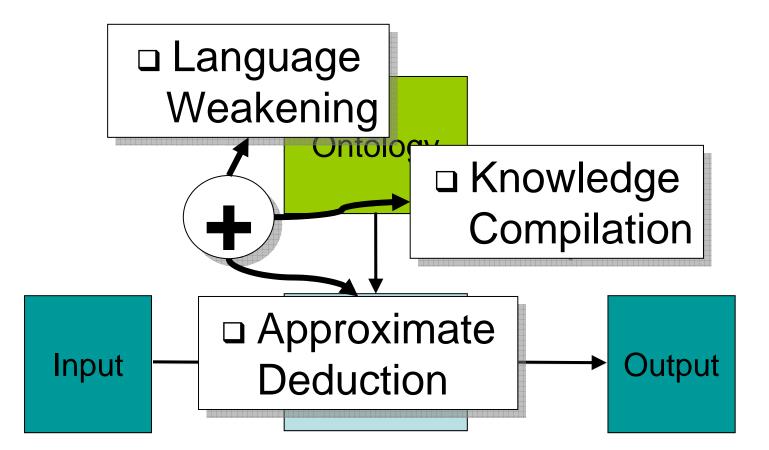


Problems tackled in KWEB





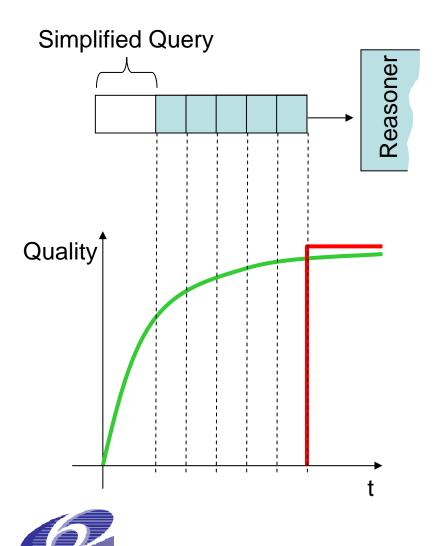








Approximate Deduction through Simplification

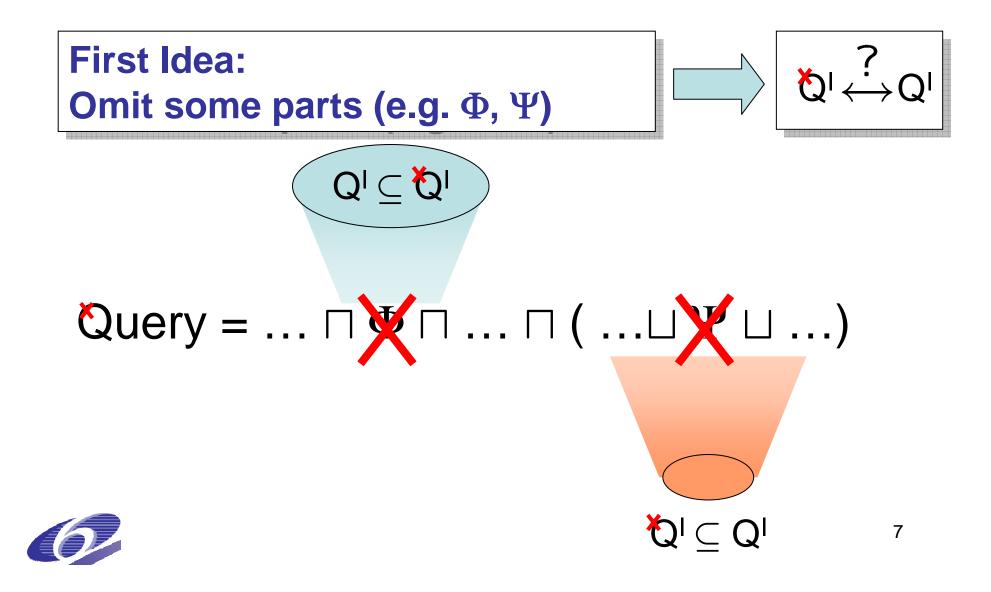


- □ Simplify query
- □ Simple query ⇒
 fast query answering
- □ Simple query ⇒ approximated answers
- Continuously complete query

□ Anytime behavior

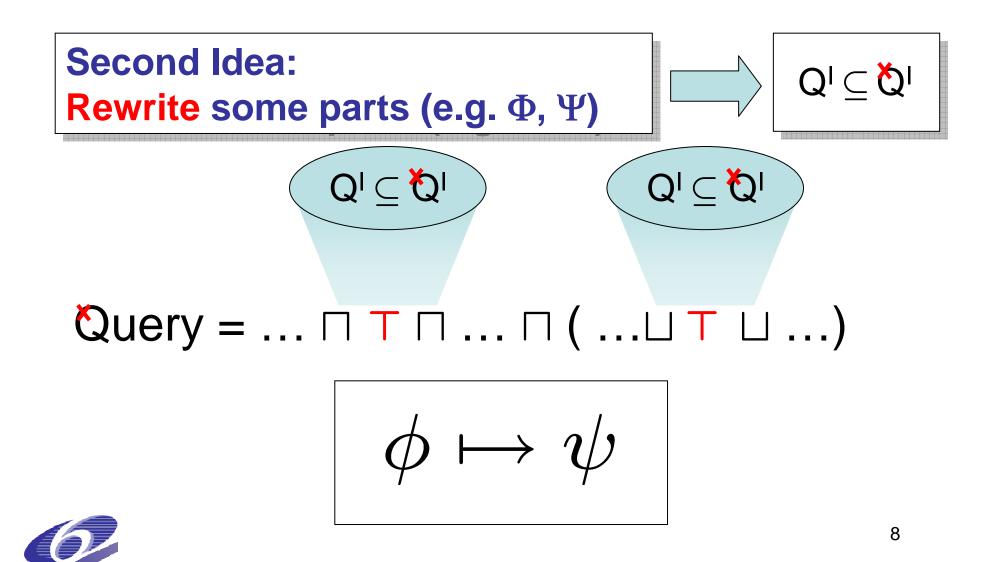


How to simplify?





How to simplify? (II)





Cadoli-Schaerf-Approximation for DLs

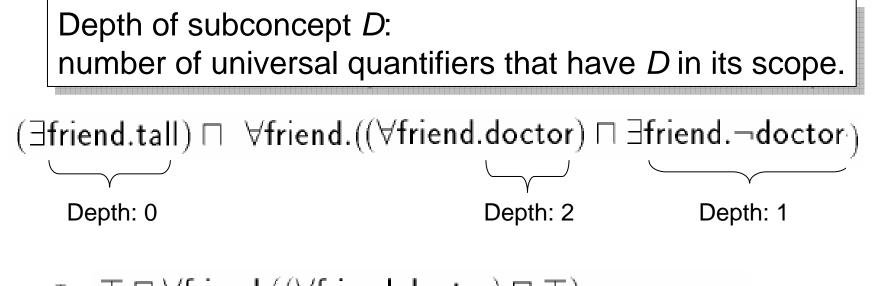
$$\begin{bmatrix} C_i^\top : \exists R.C \mapsto \top \\ C_i^\perp : \exists R.C \mapsto \bot \end{bmatrix}$$

- Replacing some sub terms in concept expressions
- Well-founded theory with (theoretically) nice results





Cadoli-Schaerf-Approximation: Example

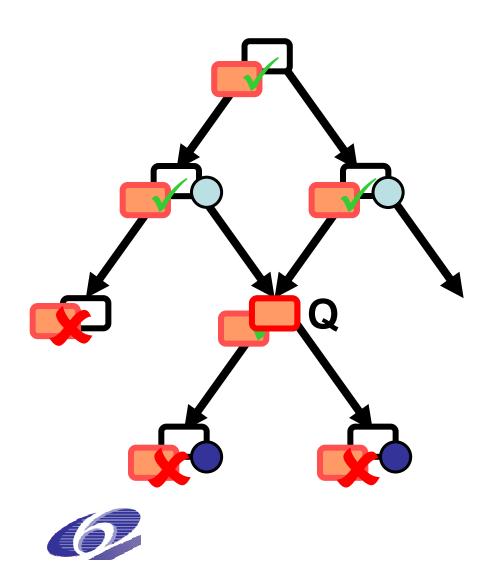


- $s_0^{\top} \top \Box \forall friend.((\forall friend.doctor) \Box \top)$
- $\mathsf{S}_1^{\scriptscriptstyle \top} \mid (\exists \mathsf{friend.tall}) \sqcap \forall \mathsf{friend.}((\forall \mathsf{friend.doctor}) \sqcap \top)$
- $\mathbf{S}_{\mathbf{2}}^{\top} \quad (\exists friend.tall) \sqcap \ \forall friend.((\forall friend.doctor) \sqcap \exists friend.\neg doctor).$





Application: Classification

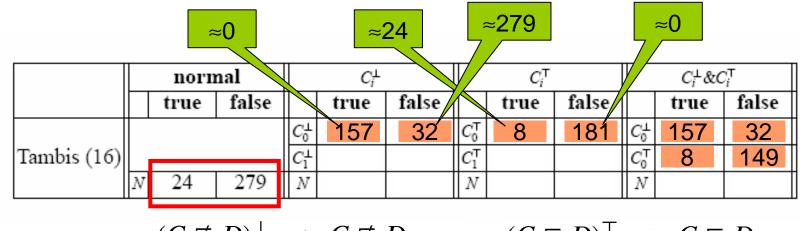


- Central process
 Classify Term Q
- Contained in
 - Generating the subsumption hierarchy
 - Instance Retrieval



Mixed Results: Classifying in TAMBIS

Application: Classification of Concepts \Rightarrow sequence of subsumption test: C \sqsubseteq D



 $(C \not\subseteq D)_i^{\perp} \Rightarrow C \not\subseteq D$

 $(C \sqsubseteq D)_i^\top \implies C \sqsubseteq D$

 \Rightarrow ($C \sqcap \neg D$) is satisfiable

 $(C \sqcap \neg D)_i^{\perp}$ is satisfiable $(C \sqcap \neg D)_i^{\top}$ is unsatisfiable \Rightarrow ($C \sqcap \neg D$) is unsatisfiable



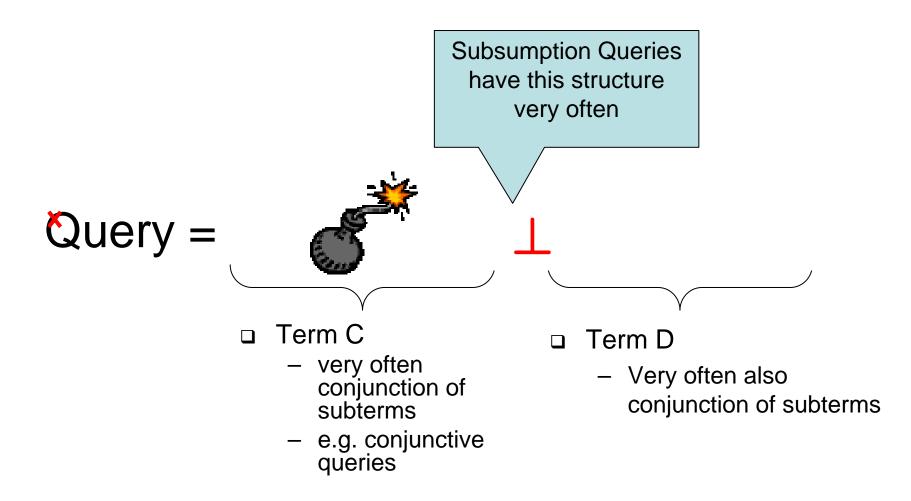


Further Results

		nor	mal	0	ч⊥ i	C	ст i	$C_i^{\perp} \& C_i^{\top}$		
		true	false	true	false	true	false	true	false	
	C_0^{\perp}	-	-	0	0	-	-	0	0	
Dolce (10)	C_0^T	-	-	-	-	0	0	0	0	
	normal	10	113	10	113	10	113	10	113	
	C_0^{\perp}	-	-	0	0	-	-	0	0	
Galen (10)	C_0^T	-	-	-	-	0	0	0	0	
	normal	10	12190	10	12190	10	12190	10	12190	
	C_0^{\perp}	-	-	0	0	-	-	0	0	
Monet (10)	C_0^T	-	-	-	-	0	0	0	0	
	normal	20	656	20	656	20	656	20	656	
	C_0^{\perp}	-	-	145	0	-	-	145	0	
MadCow (10)	C_0^T	-	-	-	-	5	140	5	140	
	normal	66	152	66	152	61	152	61	152	
	C_0^{\perp}	-	-	228	1	-	-	228	1	
Wine (10)	C_0^T	-	-	-	-	6	223	6	222	
	normal	33	252	33	251	27	252	27	251	











	normal		C_i^{\perp}			C_i^{T}				$C_i^{\perp} \& C_i^{\top}$		
	\square	true	false		true	false		true	false		true	false
				C_0^{\perp}	157	32	C_0^{T}	8	181	C_0^\perp	157	32
Tambis (16)				C_1^{\perp}	0	0	C_1^T	0	0	C_0^{T}	8	149
	N	24	279	Ν	24	247	N	16	279	Ν	16	247

Term Collapsing: 157 = 100% 65 = 35,9% 190 = 62,1%





Lessons learned

 $\phi \mapsto \psi$

Avoid Term Collapsing

– Replace ψ with something else than \top or \bot

□ Find better places to rewrite

– Ontology-adapted ϕ ?





- Find all instances a which belongs to a query Q: a:Q
- Tool InstanceStore:
 - Try to replace DL reasoning as much as possible with (scalable) DB retrieval
 - Only applyable to role-free A-Boxes $a: Q \leftrightarrow I_a \sqsubseteq Q; I_a$ concept description of instance a
- Boolean Conjunctive Queries
 - $q_1 \land \cdots \land q_n$, where q_1, \cdots, q_n are of the form *x*:*C* or (x,y):*R*
 - Restrict to those which can be converted to a concept expression C





New Approximation Method: Heuristic Ordering of Conjuncts

$$q_{1} \wedge ... \wedge q_{i} \wedge ... \wedge q_{n}$$

$$\Phi(q_{1}) \quad \Phi(q_{i}) \quad \Phi(q_{n})$$

$$\Phi(q_{n}) < \Phi(q_{1}) < \Phi(q_{i})$$

$$\boxed{q_{n} \wedge ... \wedge q_{1} \wedge q_{i}}$$

$$\boxed{Query}$$

 Compute a ranking value for each conjunct

 $\Phi(q_i): C \mapsto \mathbb{R}$

- □ Order the conjuncts q_1, \dots, q_n according to their value
- Complete approximated query with more and more expensive conjuncts





- According to the needed computation time for each conjunction
 - Estimate the computation time a priori
- According to the possible search space reduction
 - Prefer conjuncts which eliminate a lot of instances





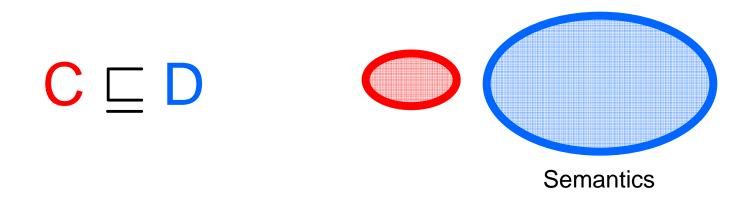
How to estimate the computation costs?

$$\begin{split} \varPhi(A) &= 1\\ \varPhi(\neg A) &= 0\\ \varPhi(C \sqcap D) &= 2 + \varPhi(C) + \varPhi(D)\\ \varPhi(C \sqcup D) &= \phi + 2 + \varPhi(C) + \varPhi(D)\\ \varPhi(\exists R.C) &= 2 + \varPhi(C)\\ \varPhi(\forall R.C) &= n + n \cdot \varPhi(C) \end{split}$$

where ϕ is the current value of $\Phi(E)$ where n is the number of existential quantifiers in E

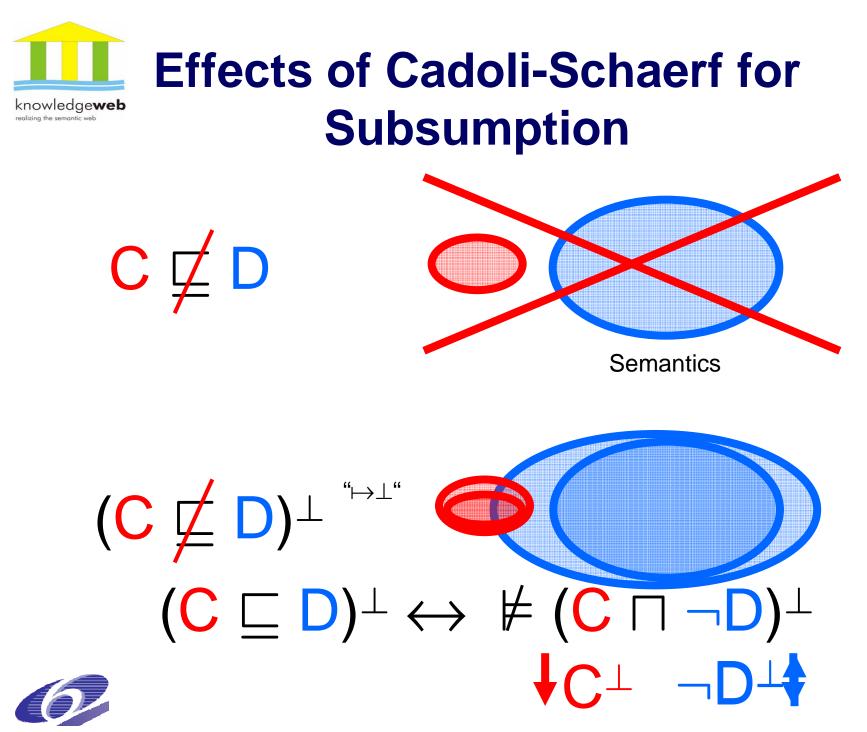






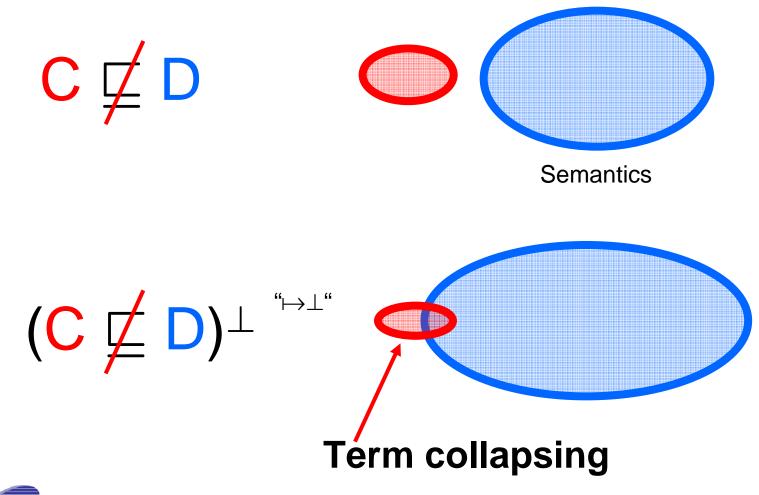
$(\mathsf{C} \sqsubseteq \mathsf{D})^{\perp} \overset{`` \mapsto \perp ``}{\hookrightarrow} \not\models \mathsf{C} \sqcap \neg \mathsf{D}$





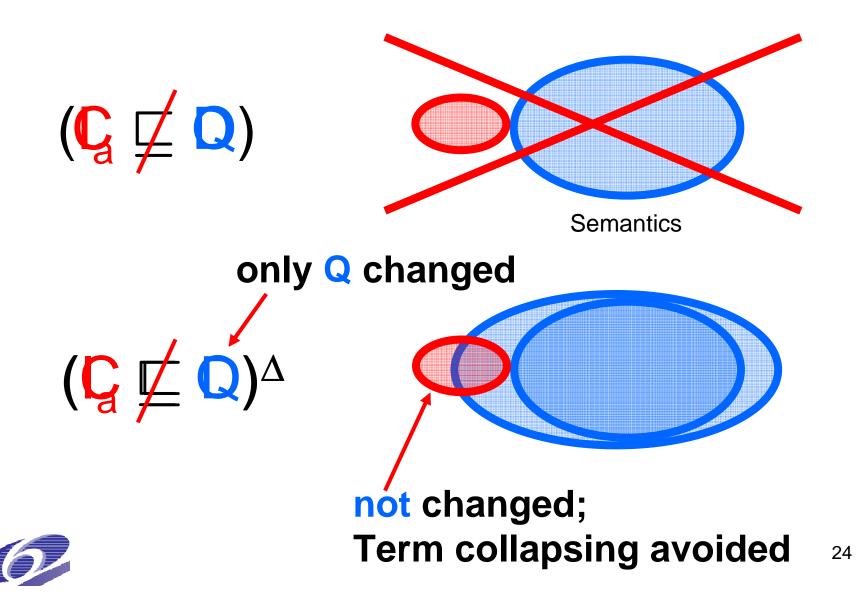


Effects of CS for Subsumption: Term Collapsing











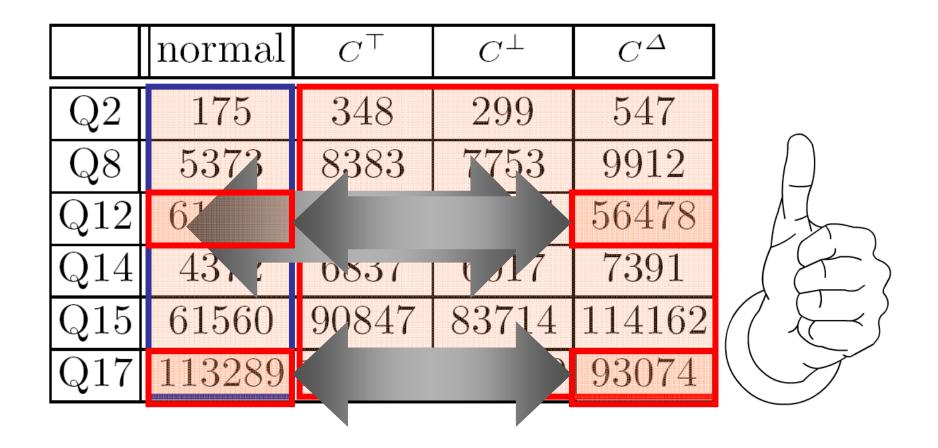
Results: Subsumption tests

	normal			C^{T}			More Levels				C^{Δ}		
		true	false		true	false		tra	A	EIL A		true	false
									\square	ण	L0	20	0
Q_2										\mathbb{T}	L1	20	0
~~~				L0		19	LO		0		L2	9	11
	$\operatorname{normal}$	9	11	$\operatorname{normal}$	9	11	normal	9	11	1	ormal	9	0
											L0		0
Q8				LO		606	LO		0		L1	10	597
	normal	10	597	$\operatorname{normal}$	10	597	normal		597	111	ormal	10	0
Q12				L0		7871		7871		Π	L0	15	856
	$\operatorname{normal}$	15	7856	$\operatorname{normal}$	15	7856	$\operatorname{normal}$	15	785	5 1	ormal	15	0
										П	L0	408	0
Q14											L1	5	403
		_		LO		407		407	0		L2	5	0
	normal	5	403	normal	5	403	normal	5	403	111	ormal	5	0
Q15				L0		6693		6693			LO		0
	$\operatorname{normal}$	46	6647	normal	46	6647	normal	46	664	7 1	ormal	46	647
Q17				L0	0	7873	L0	7873	0	Π	L0	1	872
Q11	$\operatorname{normal}$	1	7872	$\operatorname{normal}$	1	7872	$\operatorname{normal}$	1	787	2 1	ormal	1	0
												2	



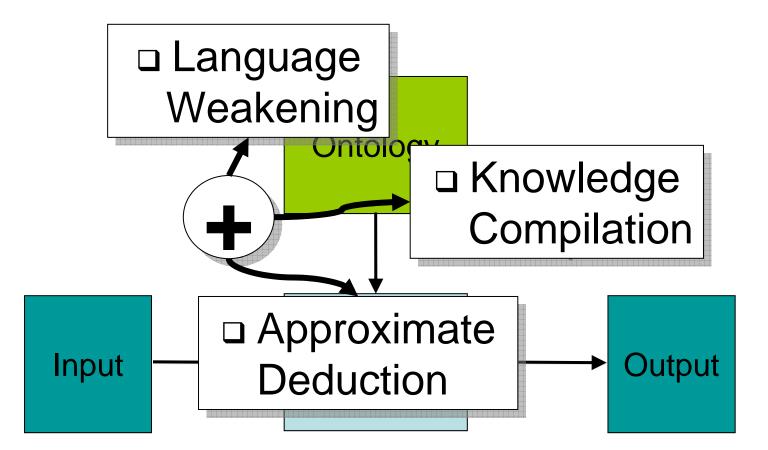


### **Results: Time**





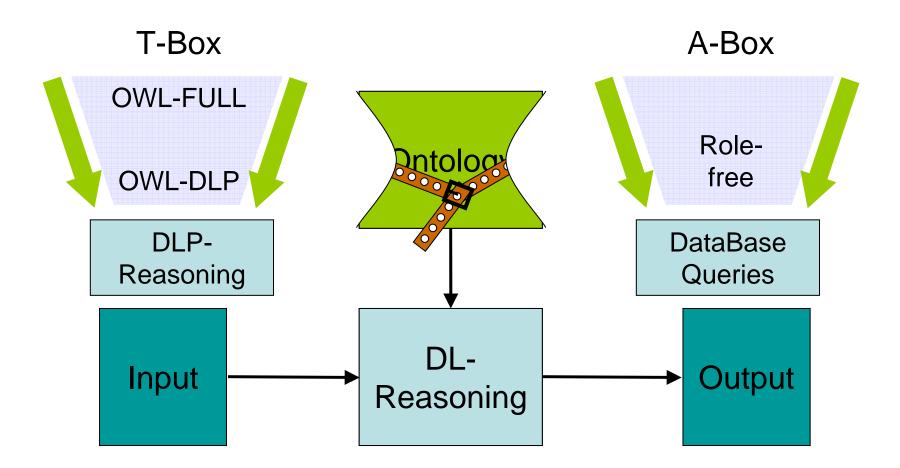






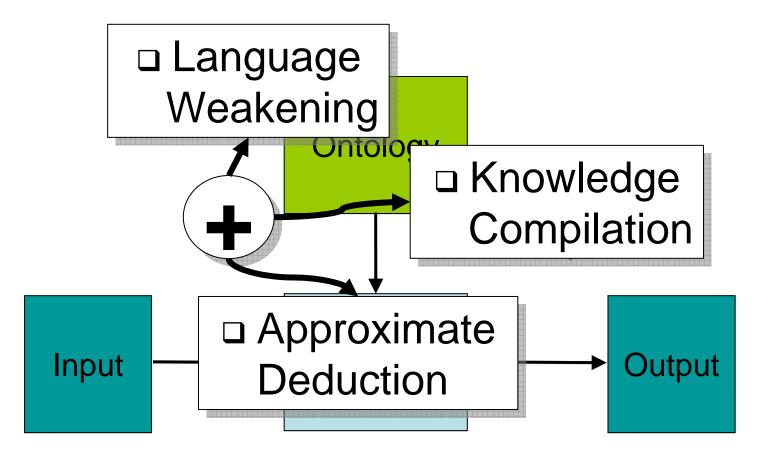


### Approximation through Language Weakening





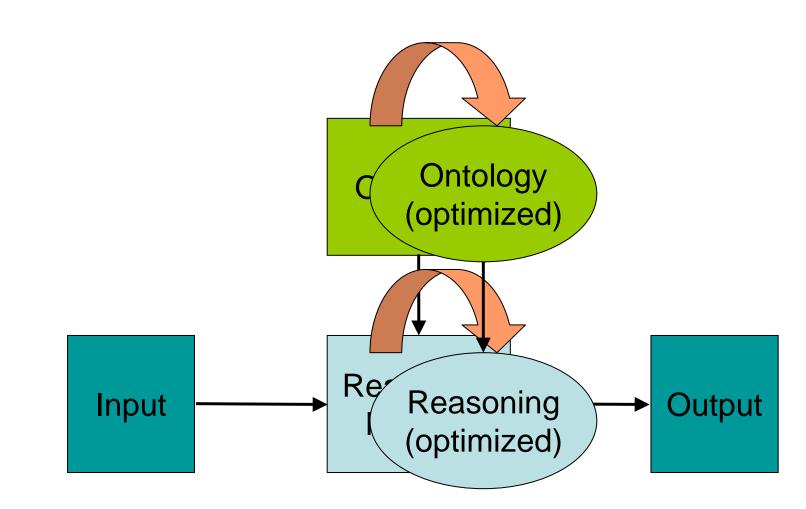








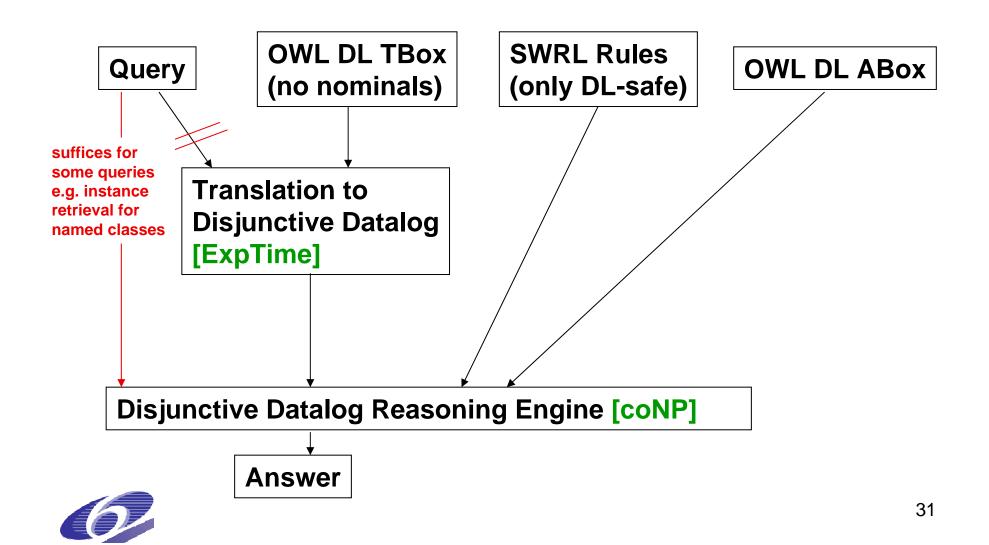
### Approximation through Knowledge Compilation

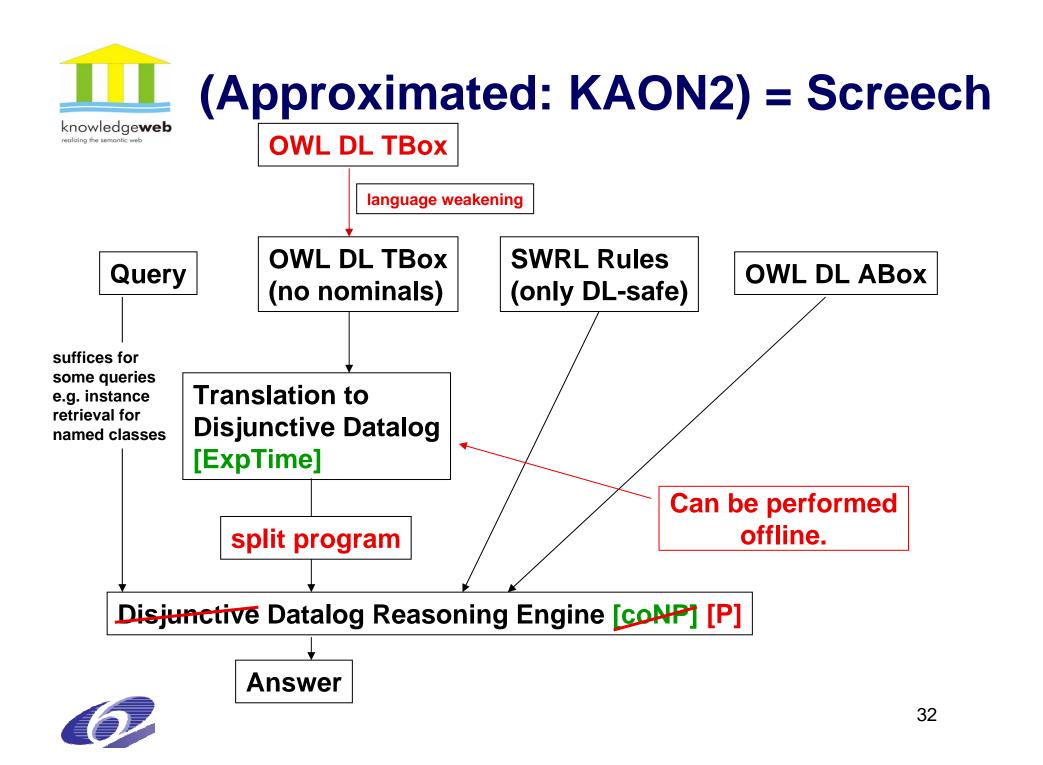






### **Standard: KAON2**







### serbian $\sqcup$ croatian $\sqsubseteq$ european eucitizen $\sqsubseteq$ european german $\sqcup$ french $\sqcup$ beneluxian $\sqsubseteq$ eucitizen **beneluxian = luxembourgian** $\sqcup$ **dutch \sqcup belgian**

serbian(ljiljana). french(julien). german(stephan). **belgian(saartje).** 

serbian(nenad). croatian(boris). croatian(denny). german(rudi). german(pascal). german(markus). indian(sudhir). german(york).





# **Screech simple example**

beneluxian  $\equiv$  luxembourgian  $\sqcup$  dutch  $\sqcup$  belgian

**KAON2** translates into the following four clauses:

Iuxembourgian(x)  $\lor$  dutch(x)  $\lor$  belgian(x)  $\leftarrow$  beneluxian(x)beneluxian(x)  $\leftarrow$  luxemburgian(x)beneluxian(x)  $\leftarrow$  dutch(x)beneluxian(x)  $\leftarrow$  belgian(x)

#### Screech split first clause:

```
\begin{array}{l} \text{luxembourgian}(x) \leftarrow \text{beneluxian}(x) \\ \text{dutch}(x) \leftarrow \text{beneluxian}(x) \\ \text{belgian}(x) \leftarrow \text{beneluxian}(x) \end{array}
```







# **Screech reasoning**

### □ data complexity is P

completebut unsound

Inference can be described in terms of standard notions from *non-monotonic reasoning* 





# Screech Performance (not optimized yet)

- □ Galen ontology
  - 673 axioms, 175 classes
  - randomly populated with 500 individuals

□ After KAON2: 267 disjunctions in 133 rules eliminated

### □ Complete run:

- queried for the extensions of all 175 Galen classes
- resulting in 5809 classifications (Screech)
  - 5353 (i.e. 92.2%) correct
- For 138 out of 175 classes: computed extension correct
- Average time saved: 39.0%







- Approximation approaches start to improve performance
  - Cadoli-Schaerf Approximation seems to not to work in practical settings
  - Heuristic approximation but performance improvements (only) in restricted cases?!
  - Screech 40% speed-up with only 8% wrong answers but only in one use-case
- □ Open questions:
  - Try to understand (theoretically) why they work
  - Benchmarking (more use-cases)
  - What about Robustness?





# Thank you for your attention!



