# Axiom Pinpointing in Description Logics

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# **Description Logics**



- Family of logic-based knowledge representation languages.
- Many DLs are decidable fragments of first-order logic.
- Close relationship to propositional modal logics.
- Design goal: good compromise between expressiveness and complexity
- Decidability and complexity results for a great variety of DLs and various inference problems, but also implementation of practical systems.
  - very expressive DLs of high worst-case complexity, but with highly optimized "practical" reasoning procedures

FacT, Racer Pellet, . . .

- inexpressive DLs with tractable inference problems, which are expressive enough for certain applications

CEL, Snorocket QuOnto, ...

• Applications: natural language processing, configuration, databases, modelling in engineering domains, ontologies (Web ontology language OWL, biomedical ontologies).



### Description logics

Constructors of the expressive DL  $\mathcal{ALCN}$ :

$$C \sqcap D, C \sqcup D, \neg C, \forall r.C, \exists r.C, (\geq n r), (\leq n r)$$

A man	$Human \sqcap \neg Female \sqcap$
that has a rich or beautiful wife	$\exists \mathit{married\_to.}(\mathit{Rich} \sqcup \mathit{Beautiful}) \sqcap$
and at least 3 children,	$(\geq 3 \ child) \ \Box$
all of whom are happy	$\forall child. Happy$

#### Axioms

concept definitions

 $Happy\_man \equiv Human \sqcap \dots$ 

General concept inclusions (GCIs)

 $Human \sqsubseteq \forall child. Human$  $\exists child. Human \sqsubseteq Tax\_Break$ 

### Inferences

**Subsumption** 

 $Happy\_man \sqsubseteq Tax\_break$ 

Satisfiability of concepts

Consistency of knowledge bases



### The inexpressive Description Logic $\mathcal{EL}$

conjunction  $C \sqcap D$ , existential restriction  $\exists r.C$ , top concept  $\top$ 



 $Frog \sqsubseteq Animal \sqcap \exists color. Green$ 

#### DL with restricted expressive power

- no value restrictions  $\forall r.C$
- can represent large biomedical ontologies: SNOMED CT, Gene Ontology, ...
- ullet  $\mathcal{EL}$  has better algorithmic properties than DLs with value restrictions



### Formal semantics

An interpretation  $\mathcal{I}$  has a domain  $\Delta^{\mathcal{I}}$  and associates

- concepts C with sets  $C^{\mathcal{I}}$ , and
- roles r with binary relations  $r^{\mathcal{I}}$ .

The semantics of the constructors is defined through identities:

- $T^{\mathcal{I}} = \Delta^{\mathcal{I}}$ ,
- $(C \sqcap D)^{\mathcal{I}} = C^{\mathcal{I}} \cap D^{\mathcal{I}}$ ,
- $(\exists r.C)^{\mathcal{I}} = \{d \mid \exists e.(d,e) \in r^{\mathcal{I}} \land e \in C^{\mathcal{I}}\}.$

The interpretation  $\mathcal{I}$  is a model of

- the general concept inclusion (GCI)  $C \sqsubseteq D$  iff  $C^{\mathcal{I}} \subseteq D^{\mathcal{I}}$ .
- the general TBox  $\mathcal{T}$  iff it satisfies all GCIs in  $\mathcal{T}$ .



### Subsumption

is concept C a subconcept of concept D?

$$\mathcal{T} \models C \sqsubseteq D \quad \text{iff} \quad C^{\mathcal{I}} \subseteq D^{\mathcal{I}}$$

for all models  $\mathcal{I}$  of  $\mathcal{T}$ 

Subsumption in  $\mathcal{EL}$  w.r.t. general TBoxes is polynomial.

- This is in strong contrast to the case of DLs with value restrictions, where subsumption w.r.t. general TBoxes is ExpTime-complete.
- Subsumption in  $\mathcal{EL}$  w.r.t. general TBoxes remains polynomial if we add the bottom concept, nominals, restricted role-value-maps, and restricted concrete domains.



### Error management and explanation

- Large ontologies often contain errors, and thus have unintended consequences.
- Even some of the intended consequences may appear to be unintuitive to users.



Understanding the reasons for unintuitive or unintended consequences can be difficult:

- In the DL version of the medical ontology SNOMED CT, the concept AmputationOfFinger is subsumed by AmputationOfHand.
- Finding the axioms that are responsible for this among the  $> 350\,000$  concept definitions in SNOMED by hand is not easy.



• Pinpointing: compute minimal subsets of the ontology that already have the consequence.

# Error management and explanation

comes in three different flavours

Pinpointing: identify the source of the consequence
 minimal subsets of the TBox from which a consequence follows
 MinAs

• Explanation: provide a convincing argument for the consequence

• Correction: provide suggestions for error resolution

maximal subsets of the TBox from which a consequence does not follow





### Pinpointing in DLs

example

$$a_1: A \sqsubseteq \exists r.A$$

$$a_2: A \sqsubseteq Y$$

$$a_4: Y \sqsubseteq B$$

$$\mathcal{T} \models A \sqsubseteq B$$

minimal axiom sets with consequence  $A \sqsubseteq B$  (MinAs):

$$\{a_2, a_4\}, \{a_1, a_2, a_3\}$$

pinpointing formula for consequence  $A \sqsubseteq B$ :

$$a_2 \wedge (a_4 \vee (a_1 \wedge a_3))$$

monotone Boolean formula whose satisfying valuations correspond to subsets that have the consequence



maximal non-axiom sets, i.e., without consequence  $A \sqsubseteq B$  (ManAs):

$${a_1, a_3, a_4}, {a_2, a_3}, {a_1, a_2}$$

## Pinpointing in DLs

equivalence of outputs

All three possible outputs (MinAs, ManAs, pinpointing formula) contain

- enough information to obtain all subsets that have the consequence
- without requiring additional DL reasoning.
- ⇒ can be transformed into each other without additional DL reasoning transformation may be exponential / require the solution of an NP-complete problem

Pinpointing formula to MinAs:

$$a_2 \wedge (a_4 \vee (a_1 \wedge a_3)) \longrightarrow \{a_2, a_4\}, \{a_1, a_2, a_3\}$$

- minimal satisfying valuations
- disjunctive normal form  $(a_2 \wedge a_4) \vee (a_1 \wedge a_2 \wedge a_3)$



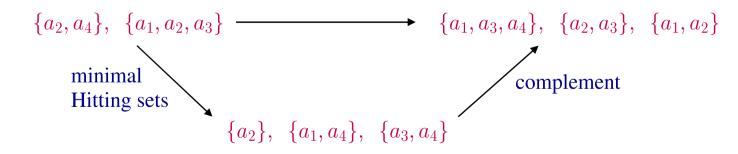
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#### MinAs to ManAs:





# Approaches to pinpointing

#### in Description Logics

#### Black Box

- employ existing inference procedure without modification:
  - + highly-optimized implementations can be reused
  - in the worst-case, the procedure needs to be invoked exponentially often
- naive approach: check for all subsets whether they have the consequence
- more sophisticated approaches work well in practice



# Approaches to pinpointing

#### in Description Logics

#### Glass Box

- modify existing inference procedure into one that directly computes minimal subsets or pinpointing formula:
  - + modified procedure is invoked only once
  - requires new implementation and optimization
- specialized approach: do this for a specific DL and a specific inference procedure
- generic approach: show how a certain class of inference procedures can be generalized to pinpointing procedures



# Glass Box Approaches

first developed by modifying tableau-based algorithms

• First introduced in [B. & Hollunder, KR'92] in the context of default reasoning in Description Logic.

Labeled version of tableau-based algorithm for the DL  $\mathcal{ALC}$  (without GCIs) to compute MinAs and ManAs: produces pinpointing formula from which both can be derived

• Re-invented in [Schlobach & Cornet, IJCAI'03] to compute minimal unsatisfiable subsets of  $\mathcal{ALC}$  TBoxes.

Labeled tableau-based algorithm similar to the one of B. & Hollunder: directly produces all MinAs

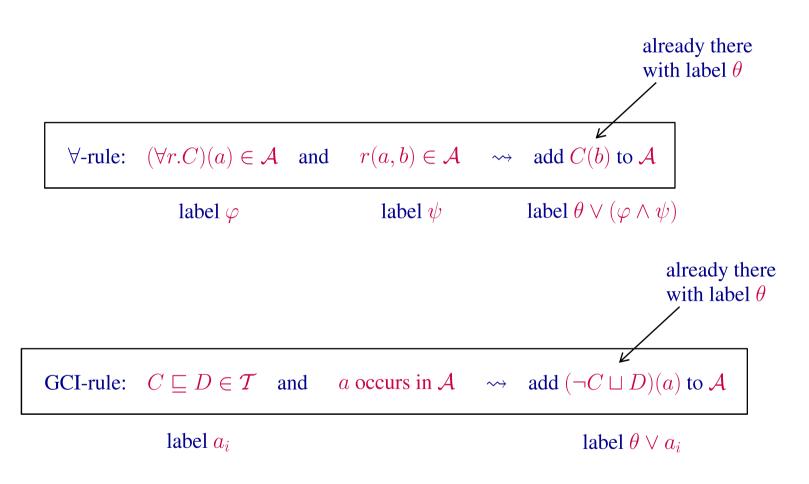
- Schlobach's approach extended in [Parsia et al., WWW'05] to more expressive DLs.
- [Lee et al., DL'06] extend approach in [B. & Hollunder, KR'92] to  $\mathcal{ALC}$  with GCIs.



• [B., et al., KI'07] introduce labeled variant of the subsumption algorithm for  $\mathcal{EL}$  with GCIs.

# Glass Box Approach

# by modifying a specific tableau-based algorithm





### Pinpointing in general tableaux

[B. & Penaloza, Tableaux'07]

[B. & Penaloza, JLC'10]

- define a general notion of a tableau system that captures
  - most of the known tableau procedures for DLs
  - also other decision procedures, like the polytime subsumption algorithm for  $\mathcal{EL}$ , congruence closure, ...
- define the pinpointing extension of a tableau system:
  - show correctness: terminating runs of the pinpointing extension compute a pinpointing formula
  - in general, termination does not transfer to the pinpointing extension
    - \* there are terminating tableau systems whose pinpointing extension does not terminate
    - \* for a given terminating tableau system, it is undecidable whether its pinpointing extension terminates
- define the notion of ordered forest tableaux:
  - always terminate and so do their pinpointing extensions



# Automata-based pinpointing

[B. & Penaloza, IJCAR'08]

[B. & Penaloza, JAR'10]

Given set of axioms T and possible consequence C, automata-based decision procedures

- construct an automaton  $\mathcal{A} = \mathcal{A}(\mathcal{T}, \mathcal{C})$ .
- perform emptiness test for A.
- $\mathcal{T} \models \mathcal{C}$  iff  $L(\mathcal{A}) = \emptyset$ .

- Define the notion of an axiomatic automaton  $\mathcal{A}(\mathcal{T},\mathcal{C})$  that "contains" all the automata  $\mathcal{A}(\mathcal{S},\mathcal{C})$  for  $\mathcal{S}\subseteq\mathcal{T}$
- Transform a given axiomatic automaton into a weighted automaton whose behaviour is a pinpoining formula



• Show how to compute the behaviour.

## Black Box Approaches

[Kalyanpur et al., ISWC'07] [B. & Suntisrivaraporn, KR-MED'08] [Suntisrivaraporn, 2009] [Horridge et al., SUM'09]

- naive approach that considers all subsets of T and tests which of them has the consequence is not practical for large ontologies like SNOMED CT (>  $360\,000$  axioms)
- more practical approaches are all based on the following idea:
  - (a) Design an efficient procedure for extracting one MinA.
  - (b) Use this procedure within Reiter's Hitting Set Tree algorithm to compute all MinAs.
- useful optimization: first compute a subset of the ontology that is
  - easy to compute
  - rather small
  - contains all MinAs



Then apply the HST approach to this subset.

### Naive linear algorithm:

- Go through the axioms according to some fixed order.
- For each axiom, check whether the consequence still holds if it is removed from the current axiom set.
- If yes, then remove it; otherwise keep it.
- Number of calls to inference procedure linear in  $|\mathcal{T}|$
- + Very simple, no overhead.

#### extracting a MinA S from a set of axioms T

### Logarithmic algorithm:

- Partition T into two halves
- For each half, check whether the consequence still holds if it is removed from the current ontology.
- If yes for one of them, then recurse on this half.
- Otherwise, do "something smart."
- + Number of calls to inference procedure logarithmic in  $|\mathcal{T}|$ , but still linear in  $|\mathcal{S}|$
- Higher overhead, which may not pay off if  $|\mathcal{T}|/|\mathcal{S}|$  is small.



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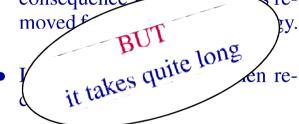
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#### extracting a MinA S from a set of axioms T

### Logarithmic algorithm:

- Partition  $\mathcal{T}$  into two halves
- For each half, check whether the consequence moved



- Otherwise, do "something smart."
- Number of calls to inference procedure logarithmic in  $|\mathcal{T}|$ , but still linear in |S|
- Higher overhead, which may not pay off if  $|\mathcal{T}|/|\mathcal{S}|$  is small.



#### experimental results for SNOMED CT

- The amputation example has exactly one MinA, which has cardinality 6.
  - The logarithmic algorithm can extract this MinA, but take 26 min.
  - First computing reachability based module and then applying linear algorithm performs much better: 0.54 sec

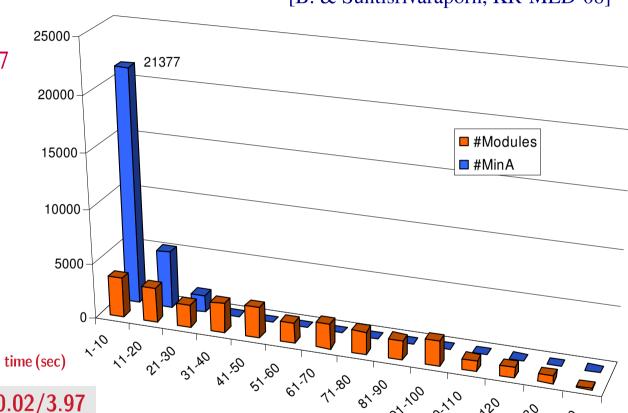
		direct-procedure-site	□ procedure-site
		AmputationOfFinger	
Ampu	tation Of	FingerWithoutThumb	■ HandExcision □
			$\exists$ roleGroup.( $\exists$ direct-procedure-site.Finger $_S \sqcap \exists$ method.Amputation)
		Amputation Of Hand	■ HandExcision □
			$\exists roleGroup.(\exists procedure\text{-site.}Hand_S \sqcap \exists method.Amputation)$
		Finger <sub>S</sub>	$\sqsubseteq$ DigitOfHand $_S$ $\sqcap$ Hand $_P$
		$Hand_P$	$\sqsubseteq Hand_S \sqcap UpperExtremity_P$



#### experimental results for SNOMED CT

[B. & Suntisrivaraporn, KR-MED'08]





extract module

0.02/3.97

logarithmic alg.

1.03/9.58

linear alg.

0.67/5.04



# Extracting ALL MinAs

# experimental results for SNOMED CT on 27 477 subsumptions

[Suntisrivaraporn, 2009]

#### Number of MinAs:

• 60% have only one MinA

• 25% have 2–9 MinAs Easy Samples

• 15% have  $\geq$  10 MinAs Hard Samples computed only the first 10

Samples	Time to extract module $\mathcal{O}_A^{\text{SNOMED}}$ (avg/max)	HST search time excl. subs. calls (avg/max)	#Subs. calls (avg/max)	Total subs. testing time (avg/max)
easy-samples	0.01 / 2.06	0.07 / 44.08	177.60 / 4732	8.80 / 131.97
hard-samples	0.02 / 3.96	0.09 / 39.90	769.98 / 4308	37.77 / 375.68



Table 6.10: Time results (second) of the modularization-based HST pinpointing algorithm on  $\mathcal{O}^{\text{SNOMED}}$ .

# Extracting ALL MinAs

experimental results for SNOMED CT on 27 477 subsumptions

[Suntisrivaraporn, 2009]

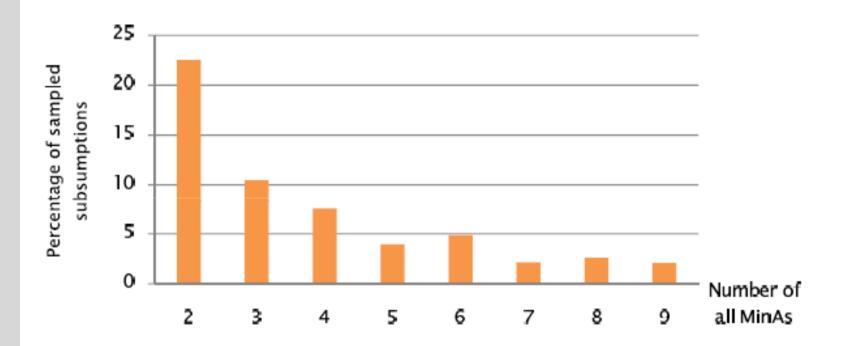


Figure 6.10: Relative frequency of the numbers of all MinAs for easy-samples in  $\mathcal{O}^{\text{SNOMED}}$ .



of pinpointing in  $\mathcal{EL}$ 

[B. et al., KI'07]

The number of MinAs can become exponential in the cardinality of  $\mathcal{T}$ :

$$\mathcal{T}_n := \{B_{i-1} \sqsubseteq P_i \sqcap Q_i, P_i \sqsubseteq B_i, Q_i \sqsubseteq B_i \mid 1 \le i \le n\}$$

$$\mathcal{T}_n \models B_0 \sqsubseteq B_n$$

- $T_n$  consists of 3n GCIs.
- The consequence  $B_0 \sqsubseteq B_n$  has  $2^n$  MinAs.



of pinpointing in  $\mathcal{EL}$ 

[B. et al., KI'07]

Determining the least cardinality of a MinA is intractable:

The following problem is NP-complete:

Given: general  $\mathcal{EL}$  TBox  $\mathcal{T}$ , concept names A, B, natural number n

Question: is there a subset  $\mathcal{T}'$  of  $\mathcal{T}$  of cardinality  $\leq n$  with  $\mathcal{T}' \models A \sqsubseteq B$ ?

Reduction from the NP-complete Hitting Set Problem:

Given: finite sets  $S_1, \ldots, S_k$ , natural number n





of pinpointing in  $\mathcal{EL}$ 

[B. et al., KI'07]

$$S_1 = \{p_{11}, \dots, p_{1\ell_1}\}, \dots, S_k = \{p_{k1}, \dots, p_{k\ell_k}\}$$

$$\mathcal{T} := \{ P_{ij} \sqsubseteq Q_i \mid 1 \le i \le k, 1 \le j \le \ell_i \} \cup \{ Q_1 \sqcap \ldots \sqcap Q_k \sqsubseteq B \} \cup \{ A \sqsubseteq P_{ij} \mid 1 \le i \le k, 1 \le j \le \ell_i \}$$

 $S_1, \ldots, S_k$  has a Hitting Set of cardinality  $\leq n$ .

iff

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There is  $\mathcal{T}' \subseteq \mathcal{T}$  of cardinality  $\leq n + k + 1$  with  $\mathcal{T}' \models A \sqsubseteq B$ .

of pinpointing in  $\mathcal{EL}$ 

[Penaloza & Sertkaya, KR'10]

Another intractable problem for MinAs: axiom relevance

Is a given axiom a possible culprit for an erroneous consequence?

Given: general  $\mathcal{EL}$  TBox  $\mathcal{T}$ , concept names  $A, B, GCI C \sqsubseteq D \in \mathcal{T}$ 

Question: is there a MinA S for  $A \sqsubseteq B$  in T such that  $C \sqsubseteq D \in S$ ?

This problem is also NP-complete!



### **Enumeration Complexity**

of pinpointing in  $\mathcal{EL}$ 

[Penaloza & Sertkaya, KR'10]

- We have seen: the number of MinAs may be exponential.
- Thus, it may take exponential time to enumerate all MinAs.
- What if the number of MinAs is actually polynomial?
   May it still take exponential time to compute them?

### Output polynomiality

An algorithm for enumerating all MinAs is output polynomial iff it runs in time polynomial in the size of the TBox and the size of all MinAs.



### **Enumeration Complexity**

of pinpointing in  $\mathcal{EL}$ 

[Penaloza & Sertkaya, KR'10]

Unless P=NP, there is no output polynomial algorithm for enumerating all MinAs in  $\mathcal{EL}$ .

This is an easy consequence of the fact that the following problem is coNP-complete:

Given: general  $\mathcal{EL}$  TBox  $\mathcal{T}$ , concept names A, B, set  $\mathcal{M}$  of subsets of  $\mathcal{T}$ .

Question: is  $\mathcal{M}$  the set of all MinAs of  $A \sqsubseteq B$  w.r.t.  $\mathcal{T}$ ?



