

**Computational Complexity
of
Description Logics:
a Friendly Introduction to Some Interesting Phenomena**

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Warm up

Which of the following subsumptions hold?

(1) r some (A and B) is subsumed by r some A
 $\exists r.(A \sqcap B) \sqsubseteq \exists r.A$

(2) (r some A) and (r only B) is subsumed by r some B
 $\exists r.A \sqcap \forall r.B \sqsubseteq \exists r.B$

(3) r only (A and not A) is subsumed by r only B
 $\forall r.(A \sqcap \neg A) \sqsubseteq \forall r.B$

(4) r some (r only A) is subsumed by r some (r some (A or not A))
 $\exists r.(\forall r.A) \sqsubseteq \exists r.(\exists r.(A \sqcup \neg A))$

(3) r only (A and B) is subsumed by (r only A) and (r only B)
 $\forall r.(A \sqcap B) \sqsubseteq \forall r.A \sqcap \forall r.B$

(4) r some B is subsumed by r only B
 $\exists r.B \sqsubseteq \forall r.A$

- we will discuss a lot of things
- but also leave out a lot
- ...please **ask** if you have a question!

Reminder: Standard DL Reasoning Problems

Given an ontology $\mathcal{O} = (\mathcal{T}, \mathcal{A})$,

- is \mathcal{O} consistent? $\mathcal{O} \models \top \sqsubseteq \perp?$
- is \mathcal{O} coherent? is there concept name A with $\mathcal{O} \models A \sqsubseteq \perp?$
- compute concept hierarchy! for all concept names A, B : $\mathcal{O} \models A \sqsubseteq B?$
- classify individuals! for all concept names A , individual names b : $\mathcal{O} \models b : B?$

Theorem 1 Let \mathcal{O} be an ontology and a an individual name **not** in \mathcal{O} . Then

1. C is satisfiable w.r.t. \mathcal{O} iff $\mathcal{O} \cup \{a : C\}$ is consistent
2. \mathcal{O} is coherent iff, for each concept name A ,
 $\mathcal{O} \cup \{a : A\}$ is consistent
3. $\mathcal{O} \models A \sqsubseteq B$ iff $\mathcal{O} \cup \{a : (A \sqcap \neg B)\}$ is **not** consistent
4. $\mathcal{O} \models b : B$ iff $\mathcal{O} \cup \{b : \neg B\}$ is **not** consistent

➔ a decision procedure for consistency decides **all** standard DL reasoning problems

Decision Procedure

- A **problem** is a set $P \subseteq M$
 - e.g., M is the set of all \mathcal{ALC} ontologies,
 - $P \subseteq M$ is the set of all **consistent** \mathcal{ALC} ontologies
 - ...and the **problem** P is to decide whether, for a given $m \in M$, we have $m \in P$
- An **algorithm** is a **decision procedure** for a problem $P \subseteq M$ if it is
 - **sound** for P : if it answers " $m \in P$ ", then $m \in P$
 - **complete** for P : if $m \in P$, then it answers " $m \in P$ "
 - **terminating**: it stops after finitely many steps on any input $m \in M$

Why does "sound and complete" not suffice for being a decision procedure?

The tableau algorithm for \mathcal{ALC} ontologies

Earlier: Anni explained a tableau algorithm for \mathcal{ALC}

Input: \mathcal{ALC} TBox \mathcal{T} , \mathcal{ALC} concept name C

Output: “yes” if C is satisfiable wrt. \mathcal{T}
“no” if not

Is this algorithm

- sound?
- complete?
- terminating?
- ...and how long does it run?

Properties of our tableau algorithm

Lemma 1: Let $\mathcal{O} = (\mathcal{T}, \mathcal{A})$ be an \mathcal{ALC} ontology in NNF. Then

1. the algorithm terminates when applied to \mathcal{T} and C
2. if the rules generate a complete & clash-free ABox, then C is satisfiable wrt. \mathcal{T}
3. if C is satisfiable wrt. \mathcal{T} , then the rules generate a clash-free & complete ABox

Corollary 1:

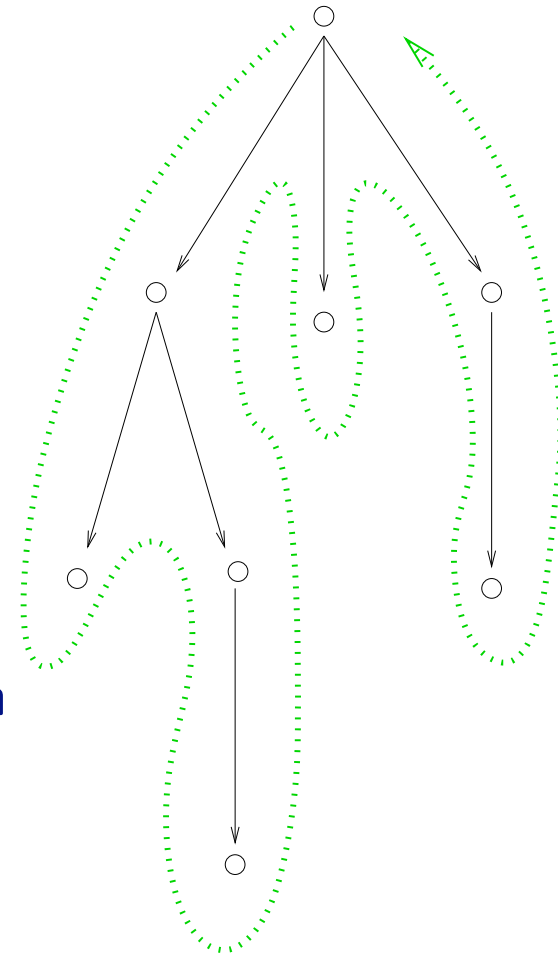
1. Our tableau algorithm **decides satisfiability** of \mathcal{ALC} concepts wrt. TBoxes.
2. Satisfiability of \mathcal{ALC} concepts (no TBox!) is decidable in **PSpace**.
3. Satisfiability of \mathcal{ALC} concepts wrt. TBoxes is decidable in **ExpSpace**.
4. \mathcal{ALC} concepts have the **finite model property**
i.e., every consistent ontology has a finite model.
5. \mathcal{ALC} concepts have the **tree model property**
i.e., every consistent ontology has a tree model.

Regarding Corollary 1.2

If we start the algorithm with $\{a: C\}$
to test satisfiability of C , and
construct ABox in non-deterministic depth-first manner
rather than constructing set of ABoxes
so that we only consider a single ABox and
re-use space for branches already visited,
mark $b: \exists R.C \in \mathcal{A}$ with “todo” or “done”

we can run tableau algorithm (even without blocking) in
polynomial space:

- ABox is of depth bounded by $|C|$, and
- we keep only a single branch in memory at any time.



Regarding Corollary 1.3

If we start the algorithm with $\{a: C\}$ and \mathcal{T} to test satisfiability of C wrt. \mathcal{T} , and construct ABox in non-deterministic depth-first manner rather than constructing set of ABoxes so that we only consider a single ABox

we can run tableau algorithm in **exponential space**:

- number of individuals in ABox is bounded by $2^{\#\text{sub}(\mathcal{T})}$

This is **not** optimal: consistency of \mathcal{ALC} ontologies is decidable in **exponential time**, in fact **ExpTime**-complete.

A tableau algorithm for \mathcal{ALC} : Summary

The tableau algorithm presented here

- **decides** consistency of \mathcal{ALC} ontologies, and thus also
- all other standard reasoning problems
- uses **blocking** to ensure termination, and
- can be implemented as such or using a **non-deterministic** alternative for the \sqcup -rule and backtracking.
- uses **P/Exp-Space**
- can be implemented in various ways,
 - order/priorities of rules
 - data structure
 - etc.
- is amenable to optimisations...

Implementing the \mathcal{ALC} Tableau Algorithm

Naive implementation of \mathcal{ALC} tableau algorithm is doomed to failure:

It constructs a

- set of ABoxes,
 - each ABox being of possibly exponential size, with possibly exponentially many individuals (see binary counting example)
 - in the presence of a GCI such as $\top \sqsubseteq (C_1 \sqcup D_1) \sqcap \dots \sqcap (C_n \sqcup D_n)$ and exponentially many individuals, algorithm might generate double exponentially many ABoxes
- ↪ requires double exponential space or
- use non-deterministic variant and backtracking to consider one ABox at a time
- ↪ requires exponential space

Implementing the \mathcal{ALC} Tableau Algorithm

Optimisations are crucial
concern every aspect of the algorithm
help in “many” cases (which?)
are implemented in various **DL** reasoners
e.g., FaCT++, Pellet, RacerPro

In the following: a selection of some vital optimisations

Optimising the \mathcal{ALC} Tableau Algorithm: Classification

Reasoners provides service “classify all concept names in \mathcal{T} ”, i.e.,

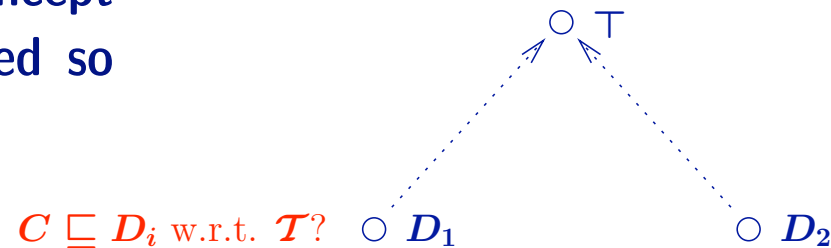
for all concept names C, D in \mathcal{T} , reasoner decides does $\mathcal{T} \models C \sqsubseteq D$?

\rightsquigarrow test consistency of $\mathcal{T} \cup \{a: (C \sqcap \neg D)\}$

$\rightsquigarrow n^2$ consistency tests!

Goal: reduce number of consistency tests when classifying TBox

Idea 1: “trickle” new concept C into hierarchy computed so far



Optimising the \mathcal{ALC} Tableau Algorithm: Classification II

Reasoners provides service “classify all concept names \mathcal{T} ”, i.e.,

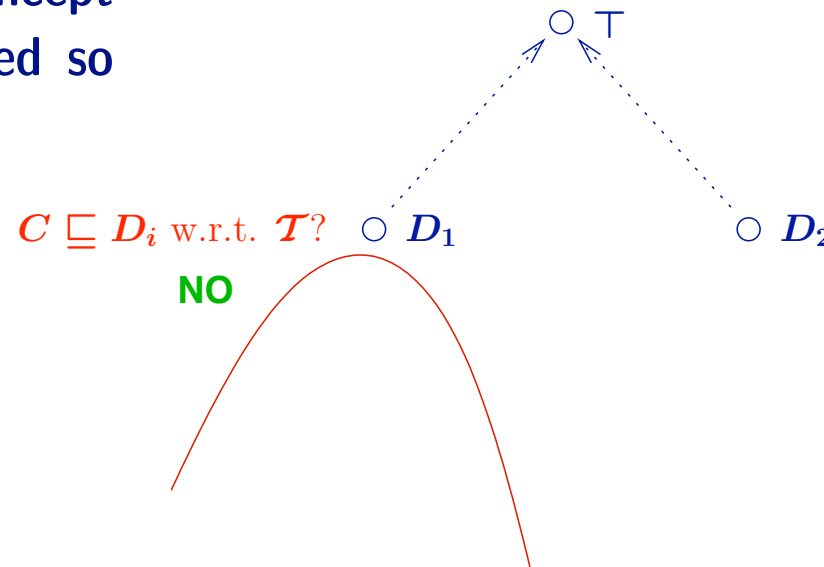
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Goal: reduce number of consistency tests when classifying TBox

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Optimising the \mathcal{ALC} Tableau Algorithm: Classification III

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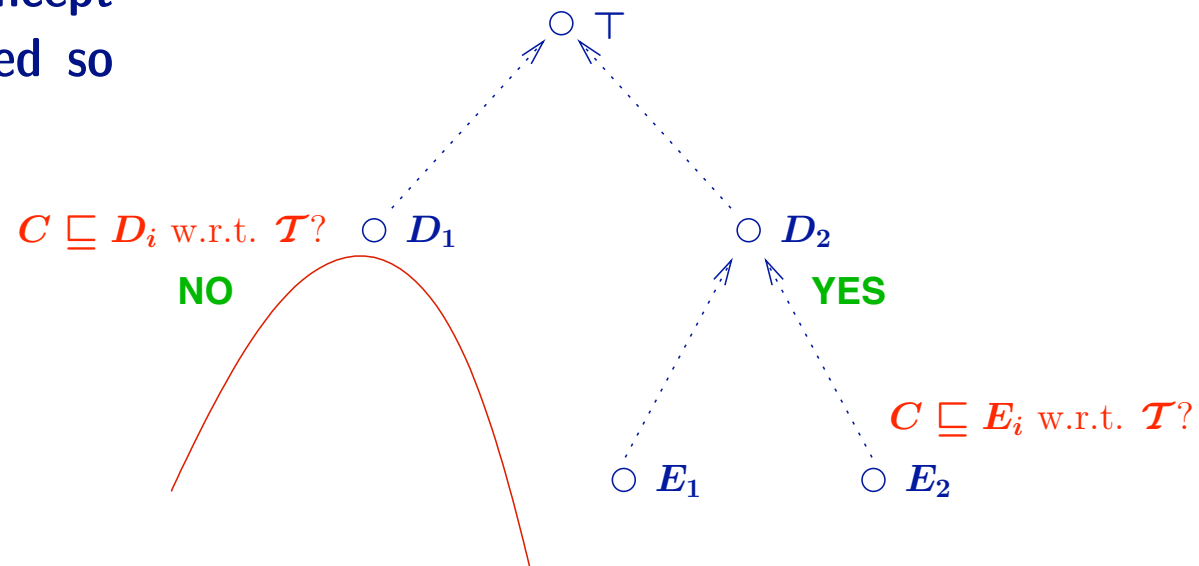
for all concept names C, D in \mathcal{T} , reasoner decides does $\mathcal{T} \models C \sqsubseteq D$?

\rightsquigarrow test consistency of $\mathcal{T} \cup \{a: (C \sqcap \neg D)\}$

$\rightsquigarrow n^2$ consistency tests!

Goal: reduce number of consistency tests when classifying TBox

Idea 1: “trickle” new concept C into hierarchy computed so far



Optimising the \mathcal{ALC} Tableau Algorithm: Classification IV

Reasoners provides service “classify all concept names \mathcal{T} ”, i.e.,

for all concept names C, D in \mathcal{T} , reasoner decides does $\mathcal{T} \models C \sqsubseteq D$?

\rightsquigarrow test consistency of $\mathcal{T} \cup \{a: (C \sqcap \neg D)\}$

$\rightsquigarrow n^2$ consistency tests!

Goal: reduce number of consistency tests when classifying TBox

Idea 2:

- maintain graph with a node for each concept name
- edges representing subsumption, disjointness ($\mathcal{T} \models A \sqsubseteq \neg B$), and non-subsumption
- initialise graph with all “obvious” information in \mathcal{T}
- to avoid testing subsumption, exploit
 - all info in ABox during tableau algorithm to update graph
 - transitivity of subsumption and its interaction with disjointness

Optimising the \mathcal{ALC} Tableau Algorithm: Absorption

Remember: for $\mathcal{T} = \{C_i \sqsubseteq D_i \mid 1 \leq i \leq n\}$,
each individual x will have n disjunctions $x: (\neg C_i \sqcup D_i)$ due to

GCI-rule: if $\mathcal{T} = \{C_i \sqsubseteq D_i \mid 1 \leq i \leq n\}$
replace \mathcal{A} with $\mathcal{A} \cup \{a: (\neg C_1 \sqcup D_1) \sqcap (\neg C_2 \sqcup D_2) \sqcap \dots \sqcap (\neg C_n \sqcup D_n)\}$

Problem: high degree of **choice** and **huge search space**
blows up set of ABoxes...we can do better:

2GCI-rule: if $C \sqsubseteq D \in \mathcal{T}$, a is not blocked, and
if C is a concept name, $a: C \in \mathcal{A}$ but $a: D \notin \mathcal{A}$,
replace \mathcal{A} with $\mathcal{A} \cup \{a: D\}$
else if $a: (\neg C \sqcup D) \notin \mathcal{A}$ for a in \mathcal{A} ,
replace \mathcal{A} with $\mathcal{A} \cup \{a: (\neg C \sqcup D)\}$

Problem: still possibly high degree of **choice** and **huge search space**...

Optimising the \mathcal{ALC} Tableau Algorithm: Absorption

Observation: many GCIs are of the form $A \sqcap \dots \sqsubseteq C$ for concept name A

e.g., $\text{Human} \sqcap \dots \sqsubseteq C$ or $\text{Device} \sqcap \dots \sqsubseteq C$

Idea: localise GCIs to concept names by transforming

$A \sqcap X \sqsubseteq C$ into equivalent $A \sqsubseteq \neg X \sqcup C$

e.g., $\text{Human} \sqcap \exists \text{owns.Pet} \sqsubseteq C$ becomes $\text{Human} \sqsubseteq \neg \exists \text{owns.Pet} \sqcup C$

For “absorbed” $\mathcal{T} = \{A_i \sqsubseteq D_i \mid 1 \leq i \leq n_1\} \cup \{C_i \sqsubseteq D_i \mid 1 \leq i \leq n_2\}$
the second, non-deterministic choice in GCI-rule is taken only n_2 times.

2GCI-rule: if $C \sqsubseteq D \in \mathcal{T}$, a is not blocked, and
 if C is a concept name, $a : C \in \mathcal{A}$ but $a : D \notin \mathcal{A}$,
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Observations: If no GCI is absorbable, nothing changes

Each absorption saves 1 disjunction per individual outside A_i ,

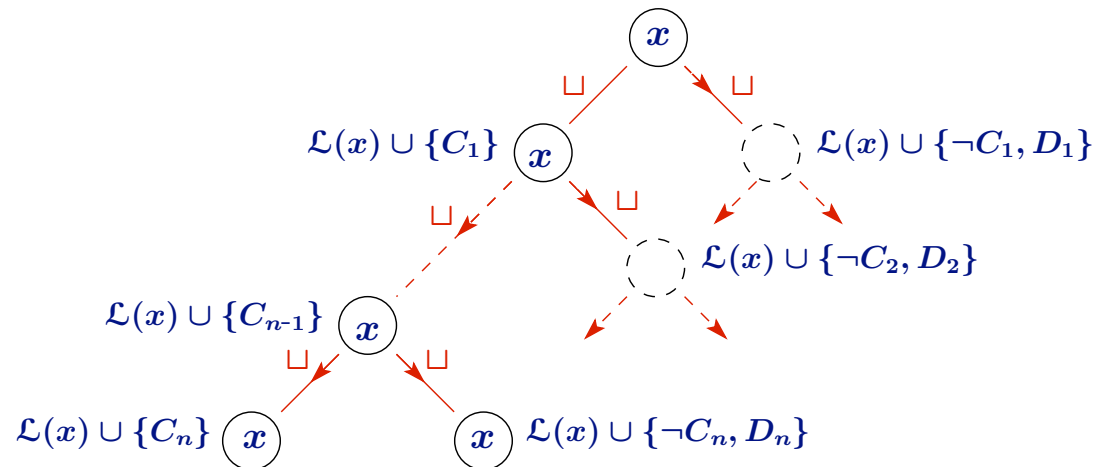
in the best case, this avoids almost all disjunctions from TBox axioms!

Optimising the \mathcal{ALC} Tableau Algorithm: Backjumping

Remember If a clash is encountered, non-deterministic algorithm backtracks

i.e., returns to last non-deterministic choice and
tries other possibility

Example $x : \exists R.(A \sqcap B) \sqcap ((C_1 \sqcup D_1) \sqcap \dots \sqcap (C_n \sqcup D_n)) \sqcap \forall R.\neg A$

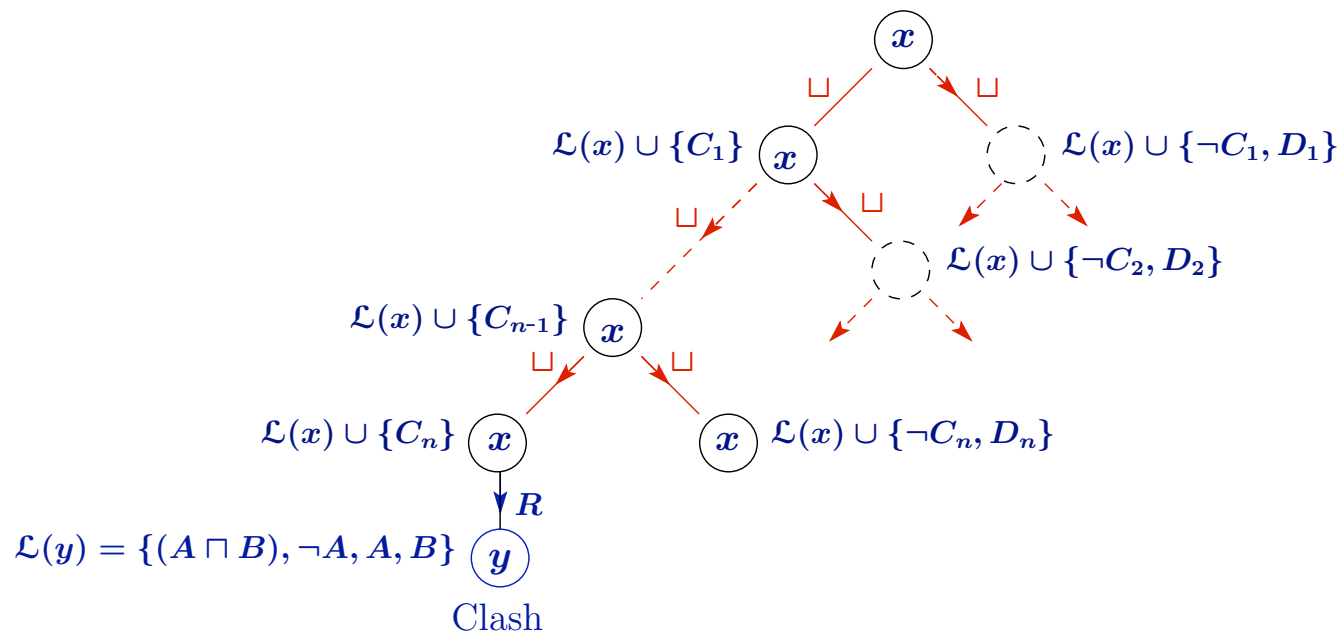


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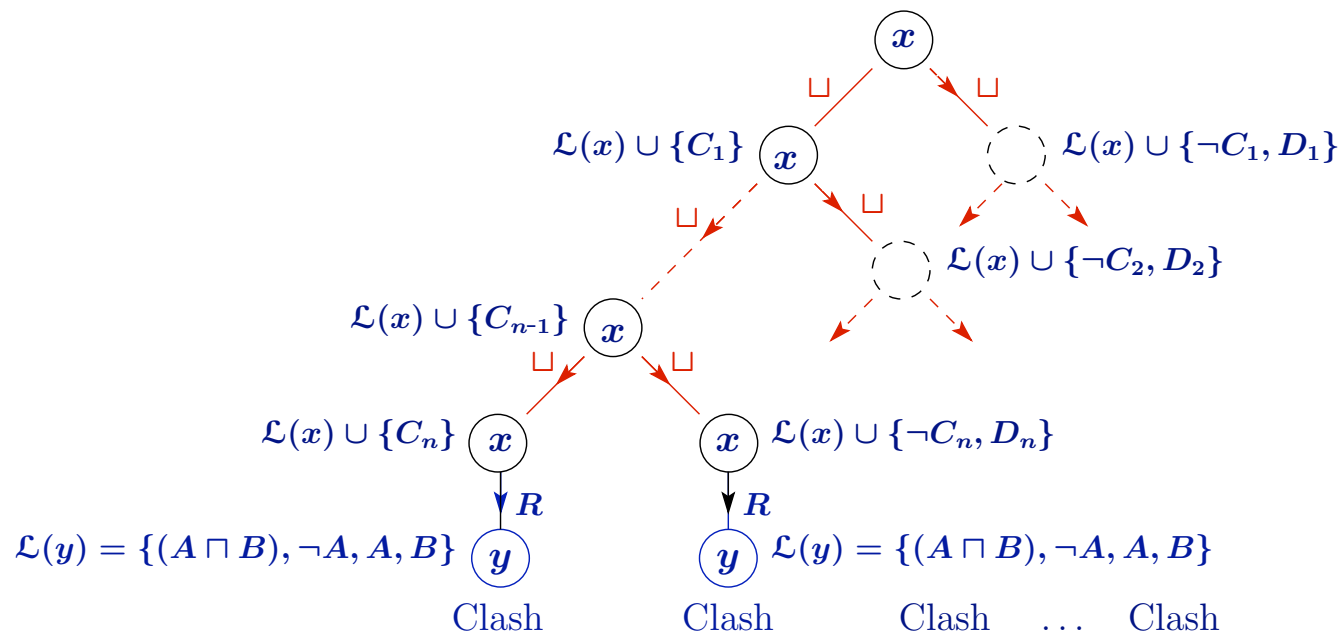


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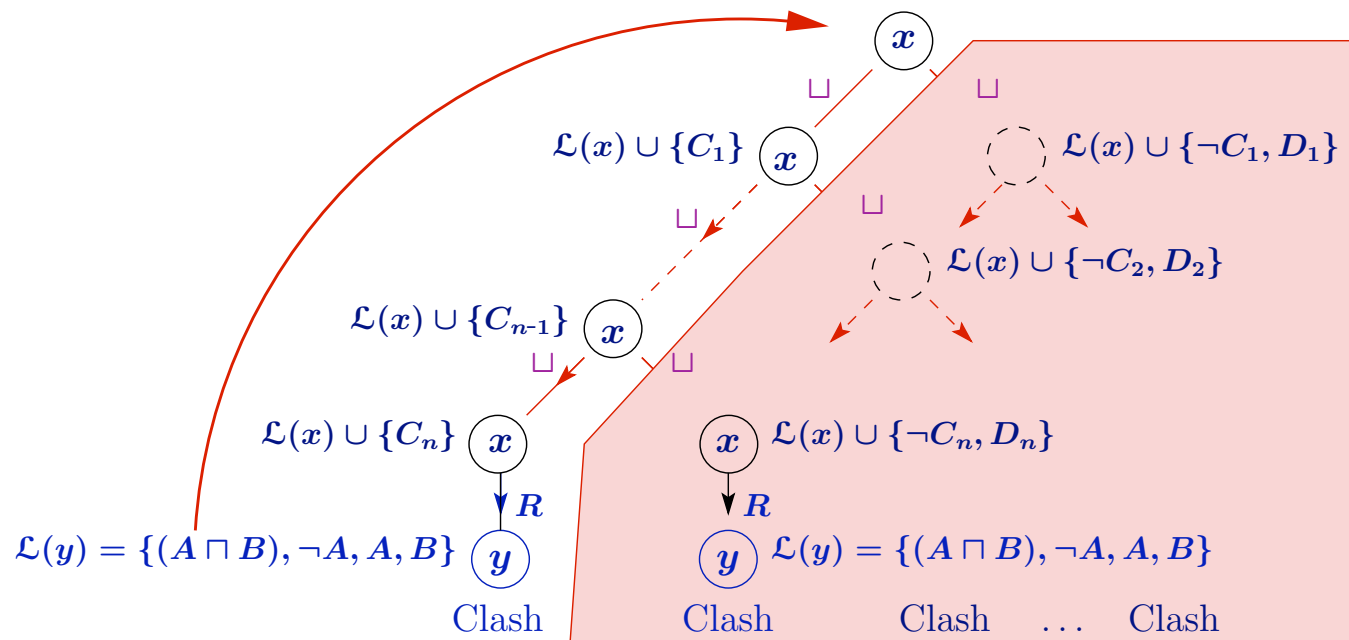


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Optimising the *ALC* Tableau Algorithm: SAT Optimisations

Finally: *ALC* extends **propositional logic**

~> heuristics developed for **SAT** are relevant

Summing up: optimisations are possible at each aspect of tableau algorithm

can dramatically enhance performance

~> do they interact?

~> how?

~> which combination works best for which “cases”?

~> is the optimised algorithm still correct?

... check out ORE 2013 results & our “Robustness” paper at DL 2013

...now for some proper computational complexity...

Computational Complexity

We have seen 1 algorithm that runs

- in PSpace without a TBox
- in non-deterministic ExpSpace with a TBox

...can we do better? How can we tell? ...perhaps try much harder, think much longer?

...how do we show that our algorithm is **optimal**? And what does that mean anyway?

↪ look at complexity...

We distinguish between

- **cognitive complexity:**

- e.g., how hard is it, for a human, to determine/understand $\mathcal{O} \models? C \sqsubseteq D$
- interesting, little understood topic
- relevant to provide tool support for ontology engineers

- **computational complexity:**

- e.g., how much time/space do we need to determine $\mathcal{O} \models? C \sqsubseteq D$
- well understood topic
- loads of results thanks to relationships DL - FOL - Modal Logic
- relevant to understand
 - * trade-off between expressivity (of a DL) and complexity of reasoning
 - * whether a given algorithm is optimal/can be improved

Computational Complexity: Decision Problems

Decision problem:

- is a subset $P \subseteq M$
- e.g., $P =$ the set of consistent \mathcal{ALC} ontologies and
 $M =$ the set of all \mathcal{ALC} ontologies
- think of it as **black box** with
 - input $m \in M$
 - output “yes” if $m \in P$
“no” if $m \notin P$

(Polynomial) reduction from $P \subseteq M$ to $P' \subseteq M'$ is a (polynomial) function π :

- $\pi : M \longrightarrow M'$
- $m \in P$ iff $\pi(m) \in P'$
- e.g., our translation $t()$ from \mathcal{ALC} to FOL
- e.g., our reduction from subsumption to ontology consistency

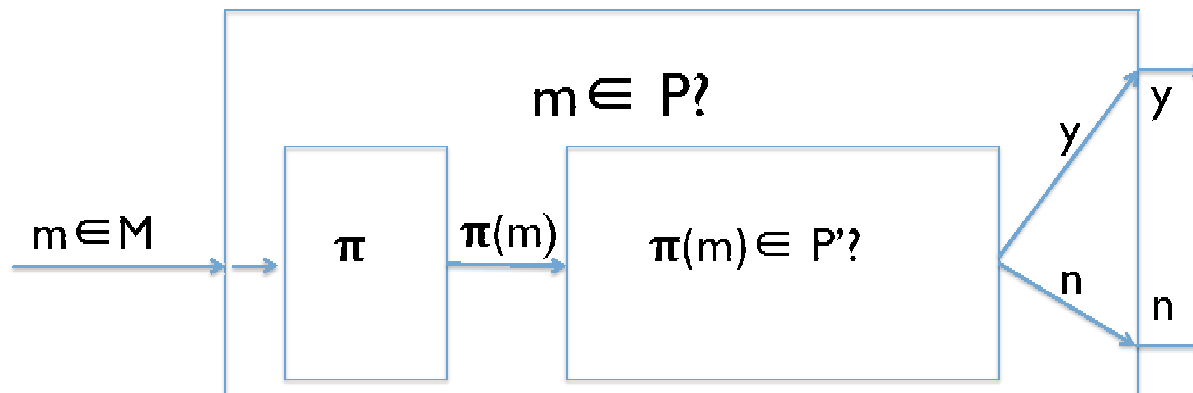
Computational Complexity: Decision Problems

Decision problem:

- is a subset $P \subseteq M$
- think of it as **black box** with
 - input $m \in M$
 - output: “yes” if $m \in P$, “no” otherwise

(Polynomial) reduction from $P \subseteq M$ to $P' \subseteq M'$ is a (polynomial) function π :

- $\pi : M \longrightarrow M'$ with $m \in P$ iff $\pi(m) \in P'$



Computational Complexity: Decision Problems

- Decision problem:**
- is a subset $P \subseteq M$
 - think of it as **black box** with
 - input $m \in M$
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(Polynomial) reduction from $P \subseteq M$ to $P' \subseteq M'$ is a (polynomial) function π :

- $\pi : M \longrightarrow M'$ with $m \in P$ iff $\pi(m) \in P'$

Fact: if $P \subseteq M$ is reducible to $P' \subseteq M'$, then P is at most as hard/complex^a as P' because P can be solved by solving P' via π

^aOf course only for suitably complex problems.

Computational Complexity

Some standard complexity classes:

Name	Meaning	Examples
L	logarithmic space	graph accessibility
P	polynomial time	model checking
NP	nondeterministic pol. time	prop. logic SAT
PSpace	polynomial space	Q-SAT
ExpTime	exponential time	
NExpTime	nondeterministic exponential time	
ExpSpace	exponential space	
...	...	
	undecidable	FOL-SAT

To determine that a problem $P \subseteq M$ is

- **in** a complexity class \mathcal{C} , it suffices to
 - design/find an algorithm
 - show that it is sound, complete, and terminating, and
 - show that this algorithm runs, for every $m \in M$, in **at most** \mathcal{C} resources
 - ...this algorithm can be a reduction to a problem known to be in \mathcal{C}
- **hard for** a complexity class \mathcal{C} , we need to
 - find a suitable problem $P' \subseteq M'$ that is **known to be hard for** \mathcal{C} and
 - a reduction $\pi(\cdot)$ from P' to P
- **complete for** a complexity class \mathcal{C} , we need to show that it is
 - in \mathcal{C} and
 - hard for \mathcal{C}

Known Complexity Results so Far:

- We have seen that *ALC* concept satisfiability (no TBox) is in **PSpace**:
 - non-deterministic tableau algorithm runs in polynomial space
 - can be extended to ABoxes
- ✓ we can't do better: *ALC* satisfiability is **PSpace-hard**:
 - but proof is a bit cumbersome
 - via a reduction of satisfiability of quantified Boolean formulae
- We have seen that *ALC* concept satisfiability w.r.t. TBoxes is in **NExpSpace**:
 - non-deterministic tableau algorithm runs in exponential space
 - can be extended to ABoxes & ontology consistency
 - can be extended to *ALCQI*, *ALCQO*, and *ALCIO*
- ✓ we can do better: *ALC* satisfiability wrt. TBoxes is **ExpTime-complete**:
 - but such (optimal) algorithm takes too long for this course
 - as is lower bound/hardness proof
(via a reduction of the halting problem of polynomial-space-bounded alternating TMs)

Worst-Case Complexity – why we bother

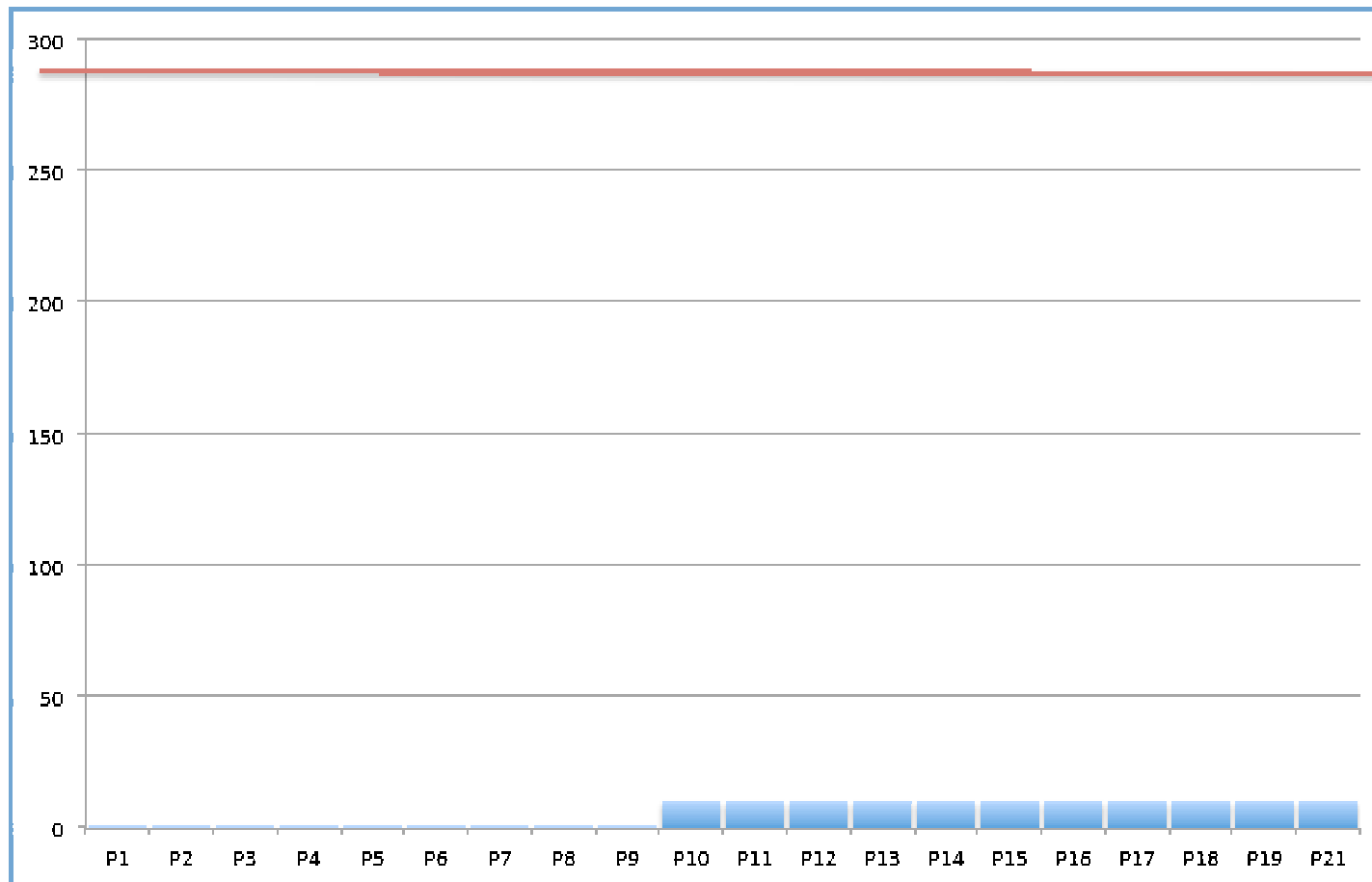
Understanding lower bounds/hardness of problems

- tell us when “in principle” improvement of algorithms is futile
- inform us of relationships of logics:
 - who is harder than who?
 - which are of similar difficulty?
- their proofs often
 - reveal interesting model theoretic properties:
 - * tree model property: each satisfiable input has a tree-shaped model
 - * finite model property: each satisfiable input has a finite model
 - use interesting translations between logics

They don't always tell us much about “typical” performance...

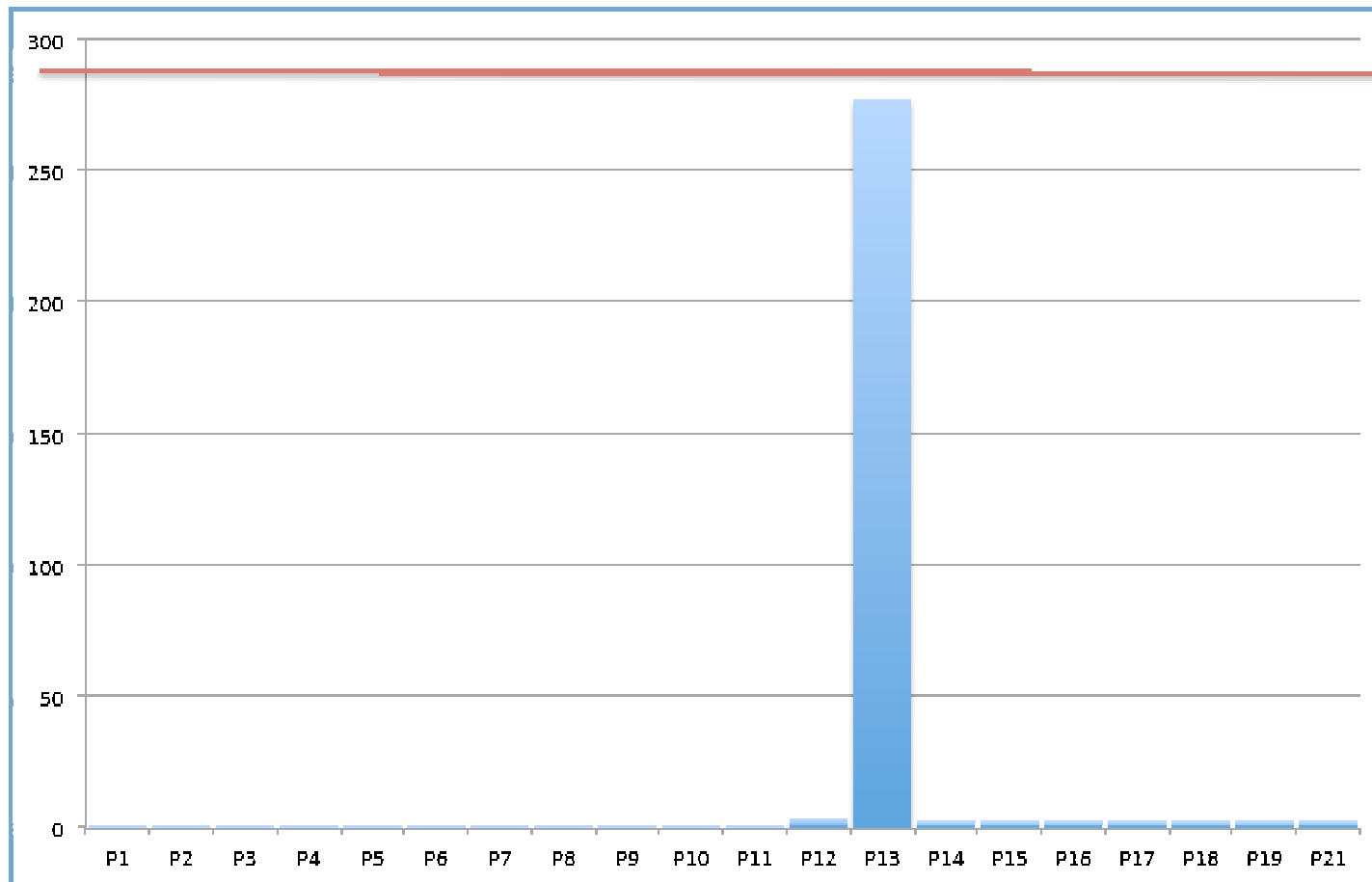
Worst-Case Complexity

Worst-case: algorithm runs, for every $m \in M$, in at most \mathcal{C} resources, e.g., like this, on all problems of size 7:



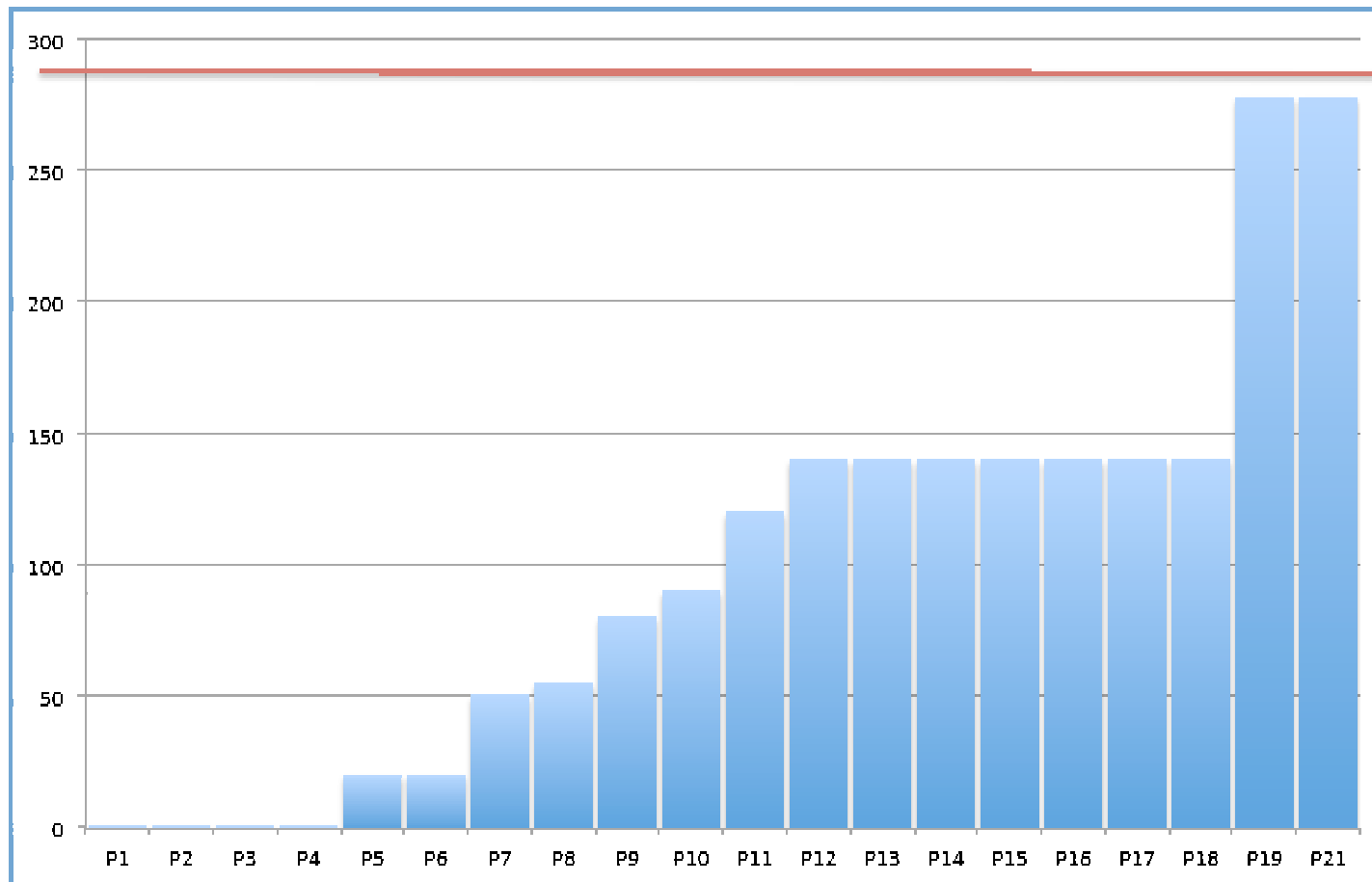
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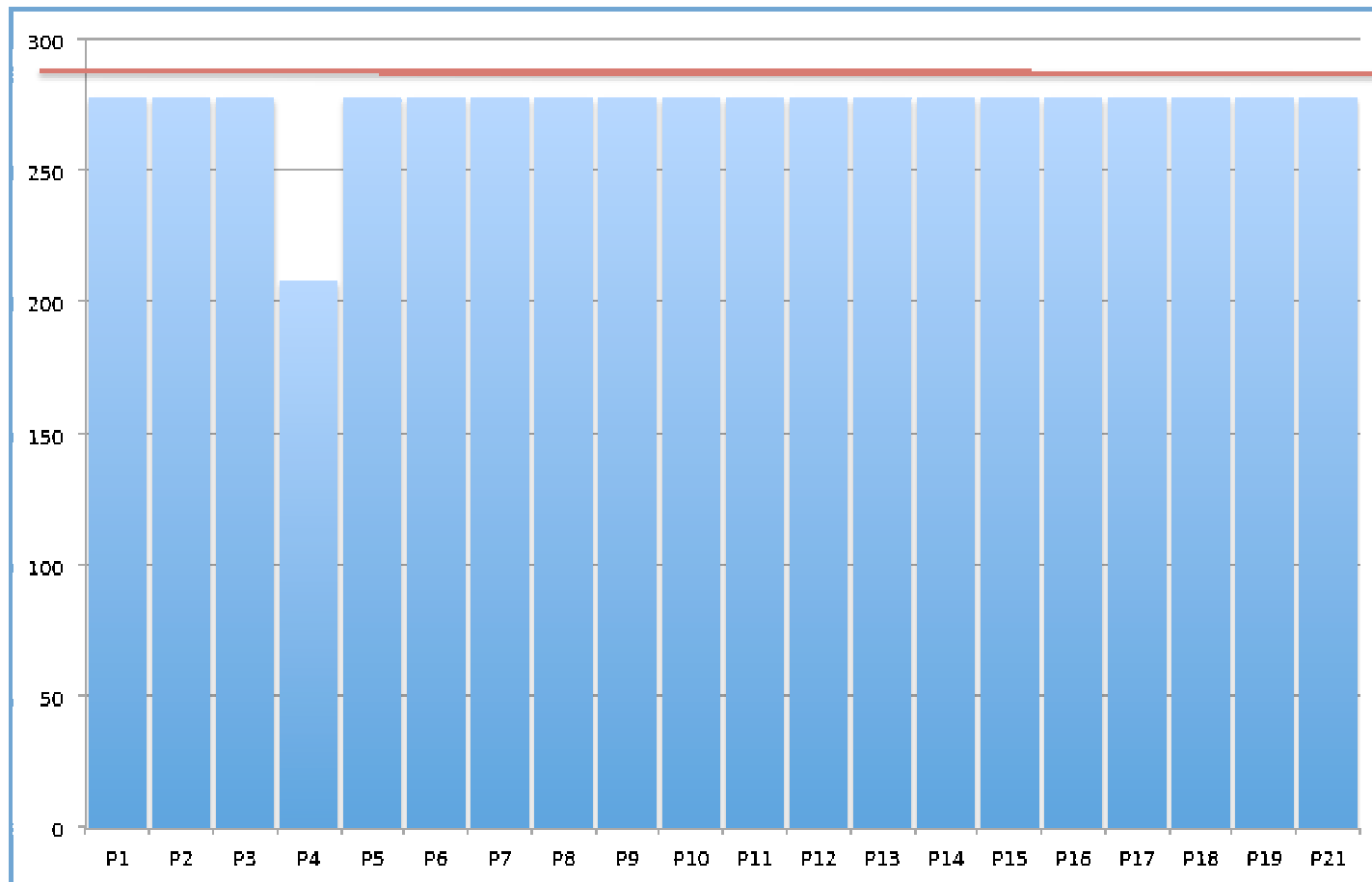
Worst-Case Complexity

Worst-case: algorithm runs, for every $m \in M$, in at most C resources, e.g., or like this, on all problems of size 7:



Worst-Case Complexity

Worst-case: algorithm runs, for every $m \in M$, in at most C resources, e.g., or like this, on all problems of size 7:



Is concept satisfiability always easier?

Earlier, we have claimed that, for \mathcal{ALC} ,

- concept satisfiability is in PSpace, but
- concept satisfiability w.r.t. a TBox is in ExpTime

Next, we will see that, for \mathcal{ALC}^u , the extension of \mathcal{ALC} with

- **universal role u with $u^{\mathcal{I}} = \Delta^{\mathcal{I}} \times \Delta^{\mathcal{I}}$**

⇒ concept satisfiability is as hard as reasoning w.r.t. a TBox, namely
ExpTime-hard

- this is typical phenomenon where the
 - certain constructors enable us to **internalise** a TBox

Internalising a TBox

Remember: for $\mathcal{T} = \{C_1 \sqsubseteq D_1, \dots, C_n \sqsubseteq D_n\}$, we use

$$C_{\mathcal{T}} = (\neg C_1 \sqcup D_1) \sqcap \dots \sqcap (\neg C_n \sqcup D_n)$$

for the **universal \mathcal{T} concept** that has to hold everywhere

Reduction: for C a concept and \mathcal{T} a TBox, define

$$\pi(C, \mathcal{T}) = C \sqcap \forall u. C_{\mathcal{T}}$$

- Lemma:**
1. C is satisfiable w.r.t. \mathcal{T} iff the concept $\pi(C, \mathcal{T})$ is satisfiable
 2. the size of $\pi(C, \mathcal{T})$ is linear in that of C plus \mathcal{T}

Corollary: satisfiability of \mathcal{ALC}^u concepts is as hard as satisfiability of \mathcal{ALC} concepts w.r.t. TBoxes is, namely ExpTime-hard

Let's do that again!

Is concept satisfiability always easier? II

Earlier, we have claimed that, for \mathcal{ALC} ,

- concept satisfiability is in PSpace, but
- concept satisfiability w.r.t. a TBox is in ExpTime

Next, we will see that, for \mathcal{ALCIO} , the extension of \mathcal{ALC} with

- **inverse roles** r^- with $(r^-)^{\mathcal{I}} = \{(y, x) \mid (x, y) \in r^{\mathcal{I}}\}$ and
- **nominals**, i.e., individual names used as concept names

⇒ concept satisfiability is as hard as reasoning w.r.t. a TBox, namely **ExpTime-hard**

- this is typical phenomenon where the
 - **combination of certain constructors enables us to internalise a TBox**

Internalising a TBox II

Remember: for $\mathcal{T} = \{C_1 \sqsubseteq D_1, \dots, C_n \sqsubseteq D_n\}$, we use

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Reduction: for C a concept and \mathcal{T} a TBox, define

$$\pi(C, \mathcal{T}) = C \sqcap C_{\mathcal{T}} \sqcap \exists p. (\{o\} \sqcap \forall p^-. (\bigsqcap_r \forall r. (\exists p. \{o\} \sqcap C_{\mathcal{T}})))$$

- Lemma:**
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Corollary: satisfiability of $\mathcal{ALC}\mathcal{IO}$ concepts is as hard as satisfiability of $\mathcal{ALC}\mathcal{IO}$ concepts w.r.t. TBoxes, namely ExpTime-hard

Are all DLs in ExpTime?

Earlier, we have claimed that $ALCQI$, $ALCQO$, and $ALCIO$ are all **ExpTime**-complete, i.e., as hard/easy as ALC

Next, we will see that consistency of $ALCQIO$ ontologies, the extension of ALC with

- **inverse roles** r^- with $(r^-)^I = \{(y, x) \mid (x, y) \in r^I\}$
- **number restrictions**, in fact functionality restrictions ($\leq 1r \top$) and
- **nominals**, i.e., individual names used as concept names

\Rightarrow is harder, namely **NExpTime**-hard

- this is typical phenomenon where
 - **combination of otherwise harmless constructors leads to increased complexity**

Domino Problems

Definition: A domino system $\mathcal{D} = (D, H, V)$

- set of domino types $D = \{D_1, \dots, D_d\}$, and
- horizontal and vertical matching conditions
 $H \subseteq D \times D$ and $V \subseteq D \times D$

A tiling for \mathcal{D} is a function:

$$t : \mathbb{N} \times \mathbb{N} \rightarrow D \text{ such that}$$
$$\langle t(m, n), t(m + 1, n) \rangle \in H \text{ and}$$
$$\langle t(m, n), t(m, n + 1) \rangle \in V$$

Domino problems: classical given \mathcal{D} , does \mathcal{D} have a tiling?

\Rightarrow well-known that this problem is undecidable [Berger66]

NexpTime given \mathcal{D} , does \mathcal{D} have a tiling for $2^n \times 2^n$ square?

\Rightarrow well-known that this problem is **NExpTime-hard**

Reduction of NExpTime Domino Problem to *ALCQIO* Consistency

To reduce the NExpTime domino problem to *ALCQIO* consistency, we need to

- define a mapping $\pi(\cdot)$ from domino problems to *ALCQIO* ontologies such that

D has an $2^n \times 2^n$ mapping iff $\pi(D)$ is consistent
and
size of $\pi(D)$ is polynomial in n

Mapping a Domino System into an *ALCQIO* Ontology

Elements in models of $\pi(D)$ will stand for points in the grid, i.e., (m, n) ...

We can express various **obligations** of the domino problem in *ALC* TBox axioms:

① each element carries **exactly one domino type** D_i

\rightsquigarrow use concept name D_i for each domino type and

$$\begin{array}{ll} \top \sqsubseteq D_1 \sqcup \dots \sqcup D_d & \% \text{ each element carries a domino type} \\ D_1 \sqsubseteq \neg D_2 \sqcap \dots \sqcap \neg D_d & \% \text{ but not more than one} \\ D_2 \sqsubseteq \neg D_3 \sqcap \dots \sqcap \neg D_d & \% \dots \\ \vdots & \\ D_{d-1} \sqsubseteq \neg D_d & \end{array}$$

Mapping a Domino System into an \mathcal{ALCQIO} Ontology

② every element has a horizontal (X -) successor and a vertical (Y -) successor

$$\top \sqsubseteq \exists X.\top \sqcap \exists Y.\top$$

③ every element satisfies D 's horizontal/vertical matching conditions:

$$\begin{array}{l}
 D_1 \sqsubseteq \bigsqcup_{(D_1,D) \in H} \forall X.D \sqcap \bigsqcup_{(D_1,D) \in V} \forall Y.D \\
 D_2 \sqsubseteq \bigsqcup_{(D_2,D) \in H} \forall X.D \sqcap \bigsqcup_{(D_2,D) \in V} \forall Y.D \\
 \vdots \\
 D_d \sqsubseteq \bigsqcup_{(D_d,D) \in H} \forall X.D \sqcap \bigsqcup_{(D_d,D) \in V} \forall Y.D
 \end{array}$$

Does this suffice?

I.e., does D have a $2^n \times 2^n$ tiling iff one D_i is satisfiable w.r.t. ① to ③?

- if yes, we have shown that satisfiability of \mathcal{ALC} is NExpTime-hard
- so no...what is missing?

Mapping a Domino System into an *ALCQIO* Ontology

Two things are missing:

1. the model must be large enough, namely $2^n \times 2^n$ and
2. for each element, its horizontal-vertical-successors **coincide** with their vertical-horizontal-successors and vice versa

This will be addressed using a “counting and binding together” trick ...

④ counting and binding together

- (a) use $A_1, \dots, A_n, B_1, \dots, B_n$ as “bits” for binary representation of **grid position**
e.g., (010, 011) is represented by an instance of $\neg A_3, A_2, \neg A_1, \neg B_3, B_2, B_1$

write GCI to ensure that X - and Y -successors are **incremented** correctly

e.g., X -successor of (010, 011) is (01**1**, 011)

e.g., Y -successor of (010, 011) is (010, **1**00)

- (b) use a nominal to ensure that there is only one (111...1, 111...1)

this implies, with $\top \sqsubseteq (\leq 1 X^- . \top) \sqcap (\leq 1 Y^- . \top)$ **uniqueness** of grid positions

④ counting and binding together

(a) \tilde{A}_i for “bit A_i is incremented correctly”:

$$\top \sqsubseteq \tilde{A}_1 \sqcap \dots \sqcap \tilde{A}_n$$

$$\tilde{A}_1 \sqsubseteq (A_1 \sqcap \forall X. \neg A_1) \sqcup (\neg A_1 \sqcap \forall X. A_1)$$

$$\begin{aligned} \tilde{A}_i \sqsubseteq & \left(\prod_{\ell < i} A_\ell \sqcap ((A_i \sqcap \forall X. \neg A_i) \sqcup (\neg A_i \sqcap \forall X. A_i)) \right) \sqcup \\ & \left(\neg \prod_{\ell < i} A_\ell \sqcap ((A_i \sqcap \forall X. A_i) \sqcup (\neg A_i \sqcap \forall X. \neg A_i)) \right) \end{aligned}$$

(add the same for the B_i s)

(b) ensure uniqueness of grid positions:

$$A_1 \sqcap \dots \sqcap A_n \sqcap B_1 \sqcap \dots \sqcap B_n \sqsubseteq \{o\} \quad \% \text{ top right } (2^n, 2^n) \text{ is unique}$$

$$\top \sqsubseteq (\leq 1 X^-. \top) \sqcap (\leq 1 Y^-. \top) \quad \% \text{ everything else is also unique}$$

Reduction of NExpTime Domino Problem to *ALCQIO* Consistency

Lemma: let $\pi(D)$ be ontology consisting of all axioms mentioned in ①-④:

- D has an $2^n \times 2^n$ tiling iff $\pi(D)$ is consistent
- size of $\pi(D)$ is polynomial (quadratic) in
 - the size of D and
 - n

Since the NExpTime-domino problem is NExpTime-hard, this implies consistency of *ALCQIO* is also NExpTime-hard:

if we **could** solve consistency of *ALCQIO* in, say, ExpTime, this would allow us to solve the domino problem also in ExpTime via $\pi(\cdot)$

Let's do this again!

Are all DLs decidable?

So far, we have extended \mathcal{ALC} with

- inverse role and
- number restrictions
- ...which resulted in logics whose reasoning problems are **decidable**
- ...we even discussed **decision procedures** for these extensions

Next, we will discuss some undecidable extension

- \mathcal{ALC} with role chain inclusions
- \mathcal{ALC} with number restrictions on complex roles

OWL 2 supports axioms of the form

- $r \sqsubseteq s$: a model of \mathcal{O} with $r \sqsubseteq s \in \mathcal{O}$ must satisfy $r^{\mathcal{I}} \subseteq s^{\mathcal{I}}$
- $\text{trans}(r)$: a model of \mathcal{O} with $\text{trans}(r) \in \mathcal{O}$ must satisfy $r^{\mathcal{I}} \circ r^{\mathcal{I}} \subseteq r^{\mathcal{I}}$,
where $p \circ q = \{(x, z) \mid \text{there is } y : (x, y) \in p \text{ and } (y, z) \in q\}$,
i.e., a model \mathcal{I} of \mathcal{O} must interpret r as a transitive relation
- $r \circ s \sqsubseteq t$: a model of \mathcal{O} with $r \circ s \sqsubseteq t \in \mathcal{O}$ must satisfy $r^{\mathcal{I}} \circ s^{\mathcal{I}} \subseteq t^{\mathcal{I}}$

subject to some complex restrictions

...why do we need restrictions?

...because axioms of this form lead to **loss of tree model property and undecidability**

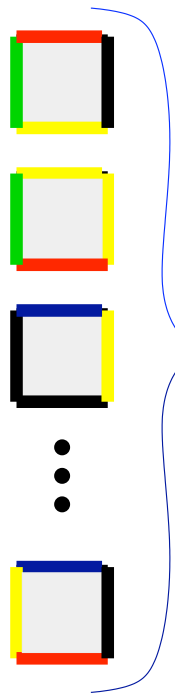
How to prove undecidability of a DL

Similar to hardness results, we prove undecidability of a DL as follows:

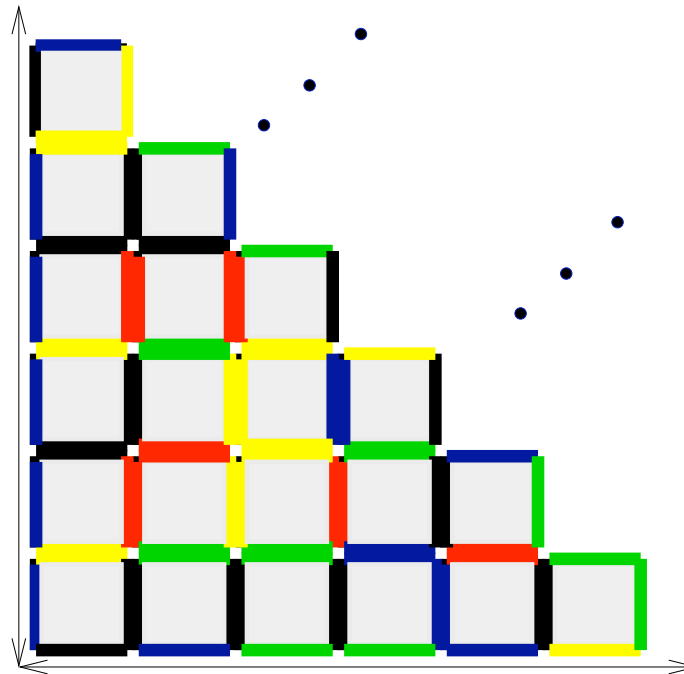
1. **fix reasoning problem**, e.g., satisfiability of a concept w.r.t. a TBox
 - remember Theorem 1?
 - if concept satisfiability w.r.t. TBox is undecidable,
 - then so is consistency of ontology
 - then so is subsumption w.r.t. TBox
 - ...
2. **pick a decision problem known to be undecidable**, e.g., the domino problem
3. **provide a (computable) mapping $\pi(\cdot)$ that**
 - takes an instance D of the domino problem and
 - turns it into a concept A_D and a TBox \mathcal{T}_D such that
 - D has a tiling if and only if A_D is satisfiable w.r.t. \mathcal{T}_D

i.e., a decision procedure of concept satisfiability w.r.t. TBoxes could be used as a decision procedure for the domino problem

The Classical Domino Problem



D ,
a fixed
set
of
dominoe
types



can we tile the
first quadrant
using D ?

The Classical Domino Problem

Definition: A domino system $\mathcal{D} = (D, H, V)$

- set of domino types $D = \{D_1, \dots, D_d\}$, and
- horizontal and vertical matching conditions $H \subseteq D \times D$ and $V \subseteq D \times D$

A tiling for \mathcal{D} is a (total) function:

$$t : \mathbb{N} \times \mathbb{N} \rightarrow D \text{ such that}$$
$$\langle t(m, n), t(m + 1, n) \rangle \in H \text{ and}$$
$$\langle t(m, n), t(m, n + 1) \rangle \in V$$

Domino problem: given \mathcal{D} , has \mathcal{D} a tiling?

It is well-known that this problem is undecidable [Berger66]

Almost Encoding the Classical Domino Problem in \mathcal{ALC}

We have already see how to express various **obligations** of the domino problem in \mathcal{ALC} TBox axioms:

① each element carries **exactly one domino type** D_i ✓

↪ use unary predicate symbol D_i for each domino type and

$$\begin{array}{ll} \top \sqsubseteq D_1 \sqcup \dots \sqcup D_d & \% \text{ each element carries a domino type} \\ D_1 \sqsubseteq \neg D_2 \sqcap \dots \sqcap \neg D_d & \% \text{ but not more than one} \\ D_2 \sqsubseteq \neg D_3 \sqcap \dots \sqcap \neg D_d & \% \dots \\ \vdots & \vdots \\ D_{d-1} \sqsubseteq \neg D_d & \end{array}$$

Almost Encoding the Classical Domino Problem in \mathcal{ALC}

- ② every element has a horizontal (X -) successor and a vertical (Y -) successor ✓

$$\top \sqsubseteq \exists X. \top \sqcap \exists Y. \top$$

- ③ every element satisfies D 's horizontal/vertical matching conditions: ✓

$$\begin{array}{l}
 D_1 \sqsubseteq \bigsqcup_{(D_1, D) \in H} \forall X. D \sqcap \bigsqcup_{(D_1, D) \in V} \forall Y. D \\
 D_2 \sqsubseteq \bigsqcup_{(D_2, D) \in H} \forall X. D \sqcap \bigsqcup_{(D_2, D) \in V} \forall Y. D \\
 \vdots \\
 D_d \sqsubseteq \bigsqcup_{(D_d, D) \in H} \forall X. D \sqcap \bigsqcup_{(D_d, D) \in V} \forall Y. D
 \end{array}$$

Does this suffice?

No, we know that it doesn't!

Encoding the Classical Domino Problem in \mathcal{ALC} with role chain inclusions

- ④ for each element, its horizontal-vertical-successors **coincide** with their vertical-horizontal-successors & vice versa

$$X \circ Y \sqsubseteq Y \circ X \text{ and } Y \circ X \sqsubseteq X \circ Y$$

Lemma: Let \mathcal{T}_D be the axioms from ① to ④.

Then \top is satisfiable w.r.t. \mathcal{T}_D iff \mathcal{D} has a tiling.

- since the domino problem is undecidable, this implies undecidability of concept satisfiability w.r.t. TBoxes of \mathcal{ALC} with role chain inclusions
- due to Theorem 1, all other standard reasoning problems are undecidable, too
- Proof: 1. show that, from a tiling for D , you can construct a model of \mathcal{T}_D
2. show that, from a model \mathcal{I} of \mathcal{T}_D , you can construct a tiling for D (tricky because elements in \mathcal{I} can have several X - or Y -successors but we can simply take **the right ones...**)

Let's do this again!

Let's do this again!

What other constructors can us help to express obligation ④?

- counting and complex roles (role chains and role intersection):

$$\top \sqsubseteq (\leq 1X.T) \sqcap (\leq 1Y.T) \sqcap (\exists(X \circ Y) \sqcap (Y \circ X).T)$$

- restricted role chain inclusions (only 1 role on RHS), and counting on **non-simple** roles:

$$\begin{aligned} \top &\sqsubseteq (\leq 1X.T) \sqcap (\leq 1Y.T) \\ X \circ Y &\sqsubseteq r \\ Y \circ X &\sqsubseteq r \\ \top &\sqsubseteq (\leq 1r.T) \end{aligned}$$

- various others...

Are all DLs hard/intractable?

Let's see a less complex DL: \mathcal{EL}

Thanks to **Thomas Schneider**, University of Bremen: he made the originals of these slides, which I borrowed and slightly modified

Are all DLs intractable?

- 1 What is \mathcal{EL} ?
- 2 Normalisation
- 3 A simple poly-time reasoning algorithm

And now . . .

- 1 What is \mathcal{EL} ?
- 2 Normalisation
- 3 A simple poly-time reasoning algorithm

Summary

\mathcal{EL} is a restriction of \mathcal{ALC} that ...

- allows only conjunction and existential restrictions
- is at the heart of OWL 2 EL
- whose standard reasoning problems are in PTime, i.e.,
i.e., there is a worst-case polynomial-time algorithm for
deciding subsumption etc.

Summary

\mathcal{EL} is a restriction of \mathcal{ALC} that ...

- allows only conjunction and existential restrictions
- is at the heart of OWL 2 EL
- whose standard reasoning problems are in PTime, i.e.,
i.e., there is a worst-case polynomial-time algorithm for deciding subsumption etc.
- can be extended to \mathcal{EL}^{++} with other features, without increase in complexity:

\perp

disjoint concepts

role (chain) inclusions

transitive roles, reflexive roles

domain and range restrictions

concept and role assertions

nominals

concrete domains

Summary

\mathcal{EL} is a restriction of \mathcal{ALC} that ...

- allows only conjunction and existential restrictions
- is at the heart of OWL 2 EL
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i.e., there is a worst-case polynomial-time algorithm for deciding subsumption etc.

- can be extended to \mathcal{EL}^{++} with other features, without increase in complexity:

\perp	domain and range restrictions
disjoint concepts	concept and role assertions
role (chain) inclusions	nominals
transitive roles, reflexive roles	concrete domains

- whose extension with inverse roles or counting increases complexity to that of \mathcal{ALC}

Syntax and semantics of \mathcal{EL}

Concepts

For C, D concepts and R a role name:

Constructor	Syntax	Example	Semantics
top	\top		$\Delta^{\mathcal{I}}$
conjunction	$C \sqcap D$	Human \sqcap Male	$C^{\mathcal{I}} \cap D^{\mathcal{I}}$
exist. restr.	$\exists r.C$	$\exists \text{hasChild.Human}$	$\{x \mid \exists y.(x, y) \in r^{\mathcal{I}} \wedge y \in C^{\mathcal{I}}\}$

Axioms

- $C \sqsubseteq D$
- $C \equiv D$ as a shortcut for “ $C \sqsubseteq D, D \sqsubseteq C$ ”

What a tiny logic !?

✓ We can say in \mathcal{EL}

Hand $\sqsubseteq \exists \text{hasPart.Finger}$

✗ but we can't say

Hand $\sqsubseteq \leq 5 \text{hasPart.Finger}$

Finger $\sqsubseteq \exists \text{hasPart}^{\neg} .\text{Hand}$

What a tiny logic !?

✓ We can say in \mathcal{EL}

Hand $\sqsubseteq \exists \text{hasPart.Finger}$

✗ We'd like to say, but can't

MildFlu \equiv Flu $\sqcap \forall \text{symptom.}\neg\text{Triv}$

Triv \equiv Cough \sqcup Sneeze. \sqcup Headache

MildFlu \equiv Flu $\sqcap \forall \text{symptom.}\neg\text{Fever}$

✗ but we can't say

Hand $\sqsubseteq \neq 5 \text{hasPart.Finger}$

Finger $\sqsubseteq \exists \text{hasPart}^{\neg}.\text{Hand}$

✓ all we can say (in \mathcal{EL}^{++}) is

MildFlu \sqsubseteq Flu

Cough \sqsubseteq Triv, Sneeze \sqsubseteq Triv, ...

MildFlu $\sqcap \exists \text{symptom.Fever} \sqsubseteq \perp$

What a tiny logic !?

✓ We can say in \mathcal{EL}

Hand $\sqsubseteq \exists \text{hasPart.Finger}$

✗ We'd like to say, but can't

MildFlu $\equiv \text{Flu} \sqcap \forall \text{symptom}.\neg \text{Triv}$

Triv $\equiv \text{Cough} \sqcup \text{Sneeze} \sqcup \text{Headache}$

MildFlu $\equiv \text{Flu} \sqcap \forall \text{symptom}.\neg \text{Fever}$

✗ but we can't say

Hand $\sqsubseteq =5 \text{hasPart.Finger}$

Finger $\sqsubseteq \exists \text{hasPart}^{\neg}.\text{Hand}$

✓ all we can say (in \mathcal{EL}^{++}) is

MildFlu $\sqsubseteq \text{Flu}$

Cough $\sqsubseteq \text{Triv}$, Sneeze $\sqsubseteq \text{Triv}$, ...

MildFlu $\sqcap \exists \text{symptom.Fever} \sqsubseteq \perp$

\mathcal{EL}^{++} is used in some large-scale ontologies, e.g., SNOMED

$\mathcal{EL}^{(+)}$ is not so tiny – an example ontology

Endocardium \sqsubseteq Tissue $\sqcap \exists \text{cont-in.HeartWall} \sqcap \exists \text{cont-in.HeartValve}$
 HeartWall \sqsubseteq BodyWall $\sqcap \exists \text{part-of.Heart}$
 HeartValve \sqsubseteq BodyValve $\sqcap \exists \text{part-of.Heart}$
 Endocarditis \sqsubseteq Inflammation $\sqcap \exists \text{has-loc.Endocardium}$
 Inflammation \sqsubseteq Disease $\sqcap \exists \text{acts-on.Tissue}$
 Heartdisease $\sqcap \exists \text{has-loc.HeartValve} \sqsubseteq$ CriticalDisease
 Heartdisease \equiv Disease $\sqcap \exists \text{has-loc.Heart}$

Taken from [Baader et al. 2006]

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Heartdisease $\sqcap \exists \text{has-loc.HeartValve} \sqsubseteq$ CriticalDisease

Heartdisease \equiv Disease $\sqcap \exists \text{has-loc.Heart}$

$\mathcal{EL}^+ \left\{ \begin{array}{l} \text{part-of} \circ \text{part-of} \sqsubseteq \text{part-of} \\ \text{part-of} \sqsubseteq \text{cont-in} \\ \text{has-loc} \circ \text{cont-in} \sqsubseteq \text{has-loc} \end{array} \right.$

Taken from [Baader et al. 2006]

Satisfiability and subsumption

Satisfiability + coherence are trivial: every \mathcal{EL} -TBox is coherent
because ?

Satisfiability and subsumption

Satisfiability + coherence are trivial: every \mathcal{EL} -TBox is coherent

- \mathcal{I} with $A^{\mathcal{I}} = \Delta^{\mathcal{I}}$ and $r^{\mathcal{I}} = \Delta^{\mathcal{I}} \times \Delta^{\mathcal{I}}$,
for all concept names A and role names r ,
satisfies every \mathcal{EL} axiom
- (\mathcal{I} with $A^{\mathcal{I}} = r^{\mathcal{I}} = \emptyset$ doesn't – **why?**)

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Subsumption ?

Satisfiability and subsumption

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for all concept names A and role names r ,
satisfies every \mathcal{EL} axiom
- (\mathcal{I} with $A^{\mathcal{I}} = r^{\mathcal{I}} = \emptyset$ doesn't – **why?**)

Subsumption isn't:

does the following TBox entail $A \sqsubseteq B$? $A' \sqsubseteq B'$?

$$\exists r.A \sqsubseteq \exists r.B$$

$$A' \equiv \exists r.\exists r.A$$

$$B' \equiv \exists r.\exists r.B$$

Satisfiability and subsumption

Satisfiability + coherence are trivial: every \mathcal{EL} -TBox is coherent

- \mathcal{I} with $A^{\mathcal{I}} = \Delta^{\mathcal{I}}$ and $r^{\mathcal{I}} = \Delta^{\mathcal{I}} \times \Delta^{\mathcal{I}}$,
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$$B' \equiv \exists r.\exists r.B$$

Without negation, they are not interreducible: Theorem 1 fails!

An Algorithm for \mathcal{EL} subsumption

Goal: present a decision procedure for subsumption in \mathcal{EL}

Outline:

- 1 Normalisation procedure
- 2 Decision procedure
(simple, naïve, without optimisations)

And now . . .

- 1 What is \mathcal{EL} ?
- 2 Normalisation
- 3 A simple poly-time reasoning algorithm

Normal form

... keeps the reasoning procedure simple

Definition

An \mathcal{EL} ontology is in **normal form** if all axioms have these forms:

$$\begin{aligned}A_1 \sqcap \dots \sqcap A_n &\sqsubseteq B \\ A &\sqsubseteq \exists r.B \\ \exists r.A &\sqsubseteq B\end{aligned}$$

$A_{(i)}, B$: concepts **names** r : role $n \geq 1$ is an integer

The normalisation procedure

- ... applies **normalisation rules** to axioms in a given TBox \mathcal{T}
- each rule transforms an axiom into one or several shorter ones
- old axiom is removed from \mathcal{T} ; new axioms are added
 \rightsquigarrow results in an “equivalent” TBox \mathcal{T}'

The normalisation rules

NF1	Input	$C \equiv D$	
	Output	$C \sqsubseteq D$	$D \sqsubseteq C$
NF2	Input	$\mathbf{C} \sqsubseteq \mathbf{D}$	
	Output	$\mathbf{C} \sqsubseteq A$	$A \sqsubseteq \mathbf{D}$

$C_{(i)}$ D arbitrary concepts
 $\mathbf{C}_{(i)}$ \mathbf{D} **complex** concepts
 B concept name
 A **fresh** concept name

The normalisation rules

NF1 Input $C \equiv D$
 Output $C \sqsubseteq D \quad D \sqsubseteq C$

NF2 Input $\mathbf{C} \sqsubseteq \mathbf{D}$
 Output $\mathbf{C} \sqsubseteq A \quad A \sqsubseteq \mathbf{D}$

NF3 Input $\exists r. \mathbf{C} \sqsubseteq D$
 Output $\mathbf{C} \sqsubseteq A \quad \exists r. A \sqsubseteq D$

NF4 Input $C_1 \sqcap \dots \sqcap \mathbf{C}_i \sqcap \dots \sqcap C_n \sqsubseteq D$
 Output $\mathbf{C}_i \sqsubseteq A \quad C_1 \sqcap \dots \sqcap A \sqcap \dots \sqcap C_n \sqsubseteq D$

$C_{(i)}$ D arbitrary concepts
 $\mathbf{C}_{(i)}$ \mathbf{D} **complex** concepts
 B concept name
 A **fresh** concept name

The normalisation rules

NF1 Input $C \equiv D$
 Output $C \sqsubseteq D \quad D \sqsubseteq C$

NF2 Input $\mathbf{C} \sqsubseteq \mathbf{D}$
 Output $\mathbf{C} \sqsubseteq A \quad A \sqsubseteq \mathbf{D}$

NF3 Input $\exists r. \mathbf{C} \sqsubseteq D$
 Output $\mathbf{C} \sqsubseteq A \quad \exists r. A \sqsubseteq D$

NF4 Input $C_1 \sqcap \dots \sqcap \mathbf{C}_i \sqcap \dots \sqcap C_n \sqsubseteq D$
 Output $\mathbf{C}_i \sqsubseteq A \quad C_1 \sqcap \dots \sqcap A \sqcap \dots \sqcap C_n \sqsubseteq D$

NF5 Input $B \sqsubseteq \exists r. \mathbf{C}$
 Output $B \sqsubseteq \exists r. A \quad A \sqsubseteq \mathbf{C}$

NF6 Input $B \sqsubseteq C_1 \sqcap \dots \sqcap C_n$
 Output $B \sqsubseteq C_1 \quad \dots \quad B \sqsubseteq C_n$

$C_{(i)}$ D arbitrary concepts
 $\mathbf{C}_{(i)}$ \mathbf{D} **complex** concepts
 B concept name
 A **fresh** concept name

The normalisation procedure

Given TBox \mathcal{T} , apply **NF1–NF7** axiom-wise until none can be applied

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Given TBox \mathcal{T} , apply NF1–NF7 axiom-wise until none can be applied

The result \mathcal{T}'

- contains new concept names A_1, \dots, A_k
- is of size linear in the size of \mathcal{T}
- is “equivalent” to \mathcal{T} ...

The normalisation procedure

Given TBox \mathcal{T} , apply NF1–NF7 axiom-wise until none can be applied

- The result \mathcal{T}'
- contains new concept names A_1, \dots, A_k
 - is of size linear in the size of \mathcal{T}
 - is “equivalent” to \mathcal{T} ...

Lemma

- For every model $\mathcal{I} \models \mathcal{T}$, there is a model $\mathcal{J} \models \mathcal{T}'$ such that $X^{\mathcal{J}} = X^{\mathcal{I}}$ for all $X \notin \{A_1, \dots, A_k\}$.
- For every model $\mathcal{J} \models \mathcal{T}'$, it holds that $\mathcal{I} \models \mathcal{T}$.

The normalisation procedure

Given TBox \mathcal{T} , apply NF1–NF7 axiom-wise until none can be applied

- The result \mathcal{T}'
- contains new concept names A_1, \dots, A_k
 - is of size linear in the size of \mathcal{T}
 - is “equivalent” to \mathcal{T} ...

Lemma

- For every model $\mathcal{I} \models \mathcal{T}$, there is a model $\mathcal{J} \models \mathcal{T}'$ such that $X^{\mathcal{J}} = X^{\mathcal{I}}$ for all $X \notin \{A_1, \dots, A_k\}$.
- For every model $\mathcal{J} \models \mathcal{T}'$, it holds that $\mathcal{I} \models \mathcal{T}$.

Consequence: \mathcal{T}' is equivalent to \mathcal{T} w.r.t. subsumption:

$$\mathcal{T} \models C \sqsubseteq D \text{ iff } \mathcal{T}' \models C \sqsubseteq D$$

for all C, D that don't use the A_i

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And now . . .

- 1 What is \mathcal{EL} ?
- 2 Normalisation
- 3 A simple poly-time reasoning algorithm**

Initial assumptions

Input: TBox \mathcal{T} , concept **names** A, B

Question: does $\mathcal{T} \models A \sqsubseteq B$ hold?

Assumption of A, B being concept *names* is no real restriction:

$$\begin{array}{c} \mathcal{T} \models C \sqsubseteq D \\ \Updownarrow \\ \mathcal{T} \cup \{A \equiv C, B \equiv D\} \models A \sqsubseteq B \end{array}$$

Deciding subsumptions via subsumer sets

Subsumer of A : a concept name B (or \top) with $\mathcal{T} \models A \sqsubseteq B$

Subsumer set $S(A)$: set that contains subsumers of A

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Representation of subsumer sets: in a labelled graph $G(\mathcal{T})$

- Nodes of $G(\mathcal{T})$ = concept names (or \top) in \mathcal{T}
- Label of node A : $S(A)$

$$B \in S(A) \quad \text{means} \quad \mathcal{T} \models A \sqsubseteq B$$
- Label of edge (A, B) : set $R(A, B)$ of roles

$$r \in R(A, B) \quad \text{means} \quad \mathcal{T} \models A \sqsubseteq \exists r.B$$

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Outline of the procedure:

- 1 Set $S(A) = \{A, \top\}$ for every A
- 2 Monotonically build $G(\mathcal{T})$
by exhaustively applying completion rules
- 3 Check whether $B \in S(A)$ to determine whether $\mathcal{T} \models A \sqsubseteq B$

The completion rules

R1 If $A_1 \sqcap \dots \sqcap A_n \sqsubseteq B \in \mathcal{T}$
 and $A_1, \dots, A_n \in S(X)$ but $B \notin S(X)$
then add B to $S(X)$

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R2 If $A \sqsubseteq \exists r.B \in \mathcal{T}$
 and $A \in S(X)$ but $r \notin R(X, B)$

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The completion rules

R1 If $A_1 \sqcap \dots \sqcap A_n \sqsubseteq B \in \mathcal{T}$
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R2 If $A \sqsubseteq \exists r.B \in \mathcal{T}$
 and $A \in S(X)$ but $r \notin R(X, B)$

 then add r to $R(X, B)$

R3 If $\exists r.A \sqsubseteq B \in \mathcal{T}$
 and $r \in R(X, Y)$ and $A \in S(Y)$ but $B \notin S(X)$

 then add B to $S(X)$

The “naïve” subsumption algorithm [Baader et al. 2006]

Algorithm 1

Input: \mathcal{EL} ontology \mathcal{T}

Output: $S(\cdot)$ such that $\mathcal{T} \models A \sqsubseteq B$ iff $B \in S(A)$

$\mathcal{T}' := \text{Normalise}(\mathcal{T})$ % by applying NF1 - NF6 exhaustively

Initialise graph for \mathcal{T}' :

For each concept name A in \mathcal{T}' (or \mathcal{T})
 create a node A with $S(A) := \{A, \top\}$
 set all edge labels $R(X, Y) := \emptyset$

Exhaustively apply rules R1-R3 to graph

Output resulting graph

Exercise

Let's apply the normalisation procedure to the TBox

$$\mathcal{T} = \left\{ \begin{array}{l} A \sqsubseteq B \sqcap \exists r.C, \\ C \sqsubseteq \exists s.D, \\ \exists r.\exists s.T \sqcap B \sqsubseteq D \end{array} \right\}$$

and then check whether it entails

$$A \sqsubseteq D.$$

Summary

Algorithm 1 ...

- terminates in time **polynomial** in the size of \mathcal{T}
- constructs a **canonical model** of \mathcal{T}
- is **sound** and **complete**: outputs *yes* iff $\mathcal{T} \models A \sqsubseteq B$
- is *one pass* (all subsumptions in 1 pass)
- is still slow for big ontologies:
...search for applicable rules over 100K concept names/nodes

Smarter versions of Algorithm 1 ...

- are goal-oriented, “one-pass”
- are implemented in the reasoners CEL, JCEL, ... for the extension \mathcal{EL}^{++}
- can be extended even to the Horn fragment of *SHIQ*

For details see [Baader et al. 2005, Baader et al. 2006, Kazakov 2009].

Bio-medical ontologies

- SNOMED, the systematized nomenclature of human and veterinary medicine

http://en.wikipedia.org/wiki/SNOMED_CT

- GALEN

<http://www.opengalen.org>

- Go, the Gene Ontology

<http://www.geneontology.org>

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Masters thesis, Technische Universität Dresden, Germany, 2005.

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What has been left out

- Loads of complexity results
- Other complexity measures
 - data complexity, relevant for OBDA – see Misha's course on Thursday!
 - average case
- Other (reasoner) performance considerations
 - what makes reasoning hard: size, **tree-width**
 - robustness
 - robustness under (small) changes to \mathcal{O} & performance homo/heterogeneity
- Other reasoning problems
 - module extraction and inseparability
 - decomposition of ontologies
 - entailment explanation and justifications

Ask us for pointers, or look at Thomas Schneider & my ESSLLI 2012 course notes

Thank you for your attention!