# Principles of <br> Knowledge Representation and Reasoning 

Nonmonotonic Reasoning

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## Introduction

## A reasoning task

- If Mary has an essay to write, she will study late in the library.
- She has an essay to write.

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- If Mary has an essay to write, she will study late in the library.
- She has an essay to write.

What do you conclude?

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## A reasoning task

- If Mary has an essay to write, she will study late in the library.
- She has an essay to write.

In empirical studies $95 \%$ of all subjects conclude (modus ponens):

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- She will study late in the library.


## A reasoning task

- If Mary has an essay to write, she will study late in the library.
- If the library is open, she will study late in the library.
- She has an essay to write.

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What do you conclude now?

## A reasoning task

- If Mary has an essay to write, she will study late in the library.
- If the library is open, she will study late in the library.
- She has an essay to write.

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In cognitive studies now only $60 \%$ of the subjects conclude:

- She will study late in the library.


## A reasoning task

- If Mary has an essay to write, she will study late in the library.
- She has an essay to write.

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Conclusion?

- She will study late in the library.

Reasoning tasks like this (suppression task; Byrne, 1989) suggest that humans often do reason as suggested by classical logics

## All logics presented so far are monotonic.

A logic is called monotonic if all (logical) conclusions from a knowledge base remain justified when new information is added to the knowledge base.

Cognitive studies indicate that everyday reasoning is often nonmonotonic (Stenning \& Lambalgen, 2008; Johnson-Laird, 2010, etc.).

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- When humans reason they use:
rules that may have exceptions:
If Mary has an essay to write, she normally will study
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default assumptions
The library is open


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- When humans reason they use:
- rules that may have exceptions:

If Mary has an essay to write, she normally will study late in the library.

- default assumptions:

The library is open.

## Defaults in knowledge bases

Often we use default assumptions when definite information is not available or when we want to fix a standard value:

1 employee(anne)
2 employee(bert)
3 employee(carla)
4 employee(detlef)
5 employee(thomas)
6 onUnpaidMPaternityLeave(thomas)
employee(X)
onUnpaidMPaternityLeave $(\mathrm{X}) \rightarrow$ gettingSalary (X) Typically: emplovee $(X) \rightarrow \neg$ onUnpaidMPaternityLeave $(X)$

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## Defaults in common sense reasoning

1 Tweety is a bird like other birds.
2 During the summer he stays in Northern Europe, in the winter he stays in Africa.

Would you expect Tweety to be able to fly? How does Tweety get from Northern Europe to Africa?

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How would you formalize this in formal logic so that you get the expected answers?

## Defaults in common sense reasoning

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- Would you expect Tweety to be able to fly?
- How does Tweety get from Northern Europe to Africa?

How would you formalize this in formal logic so that you get the expected answers?

## A formalization ...

1 bird(tweety)
2 spend-summer(tweety, northern-europe) $\wedge$ spend-winter(tweety, africa)
$3 \forall x(\operatorname{bird}(x) \rightarrow$ can-fly $(x))$
4 far-away(northern-europe, africa)
$5 \forall x y z$ (can-fly $(x) \wedge$ far-away $(y, z) \wedge$ spend-summer $(x, y) \wedge$

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Literature spend-winter $(x, z) \rightarrow$ flies $(x, y, z))$

But: The implication (3) is just a reasonable assumption. What if Tweety is an emu?

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- But: The implication (3) is just a reasonable assumption.
- What if Tweety is an emu?


## Examples of such reasoning patterns

Closed world assumption: Database of ground atoms. All ground atoms not present are assumed to be false.
Negation as failure: In PROLOG, NOT(P) means " $P$ is not provable" instead of " $P$ is provably false".
Non-strict inheritance: An attribute value is inherited only if there is no more specialized information contradicting the attribute value.

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Reasoning about actions: When reasoning about actions, it is usually assumed that a property changes only if it has to change, i.e., properties by default do not change.

## Default, defeasible, and nonmonotonic reasoning

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Default reasoning: Jump to a conclusion if there is no information that contradicts the conclusion.

Defeasible reasoning: Reasoning based on assumptions that can turn out to be wrong: conclusions are defeasible. In particular, default reasoning is defeasible.

Nonmonotonic reasoning: In classical logic, the set of consequences grows monotonically with the set of premises. If reasoning is defeasible, then reasoning becomes nonmonotonic.

## Approaches to nonmonotonic reasoning

- Consistency-based: Extend classical theory by rules that test whether an assumption is consistent with existing beliefs
$\Rightarrow$ Nonmonotonic logics such as DL (default logic), NMLP (nonmonotonic logic programming)

Entailment-based on normal models: Models are ordered
Different forms of

## Approaches to nonmonotonic reasoning

- Consistency-based: Extend classical theory by rules that test whether an assumption is consistent with existing beliefs
$\Rightarrow$ Nonmonotonic logics such as DL (default logic), NMLP (nonmonotonic logic programming)

■ Entailment-based on normal models: Models are ordered by normality. Entailment is determined by considering the most normal models only.
$\Rightarrow$ Circumscription, preferential and cumulative logics

## NM Logic - Consistency-based

If $\varphi$ typically implies $\psi, \varphi$ is given, and it is consistent to assume $\psi$, then conclude $\psi$.


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$\Rightarrow$ can-fly(tweety)

5 ... + emu(tweety)
$\Rightarrow$ can-fly(tweety)

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## NM Logic - Normal models

If $\varphi$ typically implies $\psi$, then the models satisfying $\varphi \wedge \psi$ should be more normal than those satisfying $\varphi \wedge \neg \psi$.

Similar idea: try to minimize the interpretation of "Abnormality" predicates.

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Minimize interpretation of Ab :
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## Default Logic - Outline

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Special Kinds of Defaults

## Reiter's default logic: motivation

- We want to express something like "typically birds fly".
- Add non-logical inference rule

$$
\frac{\operatorname{bird}(x): \operatorname{can}-f l y(x)}{\operatorname{can}-\operatorname{fly}(x)}
$$

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$$
\begin{aligned}
& \forall x(\text { penguin }(x) \rightarrow \neg \text { can-fly }(x)) \\
& \forall x(\text { emu }(x) \rightarrow \neg \text { can-fly }(x)) \\
& \forall x(\operatorname{kiwi}(x) \rightarrow \neg \text { can-fly }(x))
\end{aligned}
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## Reiter's default logic: motivation

- We want to express something like "typically birds fly".
- Add non-logical inference rule

$$
\frac{\operatorname{bird}(x): \text { can-fly }(x)}{\operatorname{can}-\operatorname{fly}(x)}
$$

with the intended meaning:
If $x$ is a bird and if it is consistent to assume that $x$ can fly, then conclude that $x$ can fly.

- Exceptions can be represented as formulae:

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## Formal framework

- FOL with classical provability relation $\vdash$ and deductive closure: $\operatorname{Th}(\Phi):=\{\varphi \mid \Phi \vdash \varphi\}$

Default Logic
Basics


Prerequisite: must have been derived before rule can be applied.
Consistency condition: the negation may not be derivable. Consequence: will be concluded.

A default rule is closed if it does not contain free variables.
(Closed) default theory: A pair $\langle D, W\rangle$, where $D$ is a
countable set of (closed) default rules and $W$ is a countable set of FOL formulae.

## Formal framework

■ FOL with classical provability relation $\vdash$ and deductive closure: $\operatorname{Th}(\Phi):=\{\varphi \mid \Phi \vdash \varphi\}$

- Default rules: $\frac{\alpha: \beta}{\gamma}$
$\alpha$ : Prerequisite: must have been derived before rule can be applied.
$\beta$ : Consistency condition: the negation may not be derivable.
$\gamma$ : Consequence: will be concluded.

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- A default rule is closed if it does not contain free variables.
- (Closed) default theory: A pair $\langle D, W\rangle$, where $D$ is a countable set of (closed) default rules and $W$ is a countable set of FOL formulae.


## Extensions of default theories

Default theories extend the theory given by $W$ using the default rules in $D(\rightsquigarrow$ extensions). There may be zero, one, or many extensions.

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One extension contains $b$, the other contains $c$.
Intuitively, an extension is a set of beliefs resulting from $W$ and $D$.

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Default theories extend the theory given by $W$ using the default rules in $D(\rightsquigarrow$ extensions). There may be zero, one, or many extensions.

## Example

$$
\begin{aligned}
W & =\{a, \neg b \vee \neg c\} \\
D & =\left\{\frac{a: b}{b}, \frac{a: c}{c}\right\}
\end{aligned}
$$

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One extension contains $b$, the other contains $c$.
Intuitively, an extension is a set of beliefs resulting from $W$ and $D$.

## Decision problems about extensions in default logic

Existence of extensions: Does a default theory have an extension?
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## Default Logic

# Basics 

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Credulous reasoning: If $\varphi$ is in at least one extension, $\varphi$ is a credulous default conclusion.
Skeptical reasoning: If $\varphi$ is in all extensions, $\varphi$ is a skeptical default conclusion.

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## Extensions (informally)

Desirable properties of an extension $E$ of $\langle D, W\rangle$ :

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Further requirement: Application of default rules must follow in sequence (aroundedness).

## Extensions (informally)

Desirable properties of an extension $E$ of $\langle D, W\rangle$ :

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then $\gamma \in E$.

- Further requirement: Application of default rules must follow in sequence (groundedness).


## Groundedness

## Example

$$
\begin{aligned}
W & =\emptyset \\
D & =\left\{\frac{a: b}{b}, \frac{b: a}{a}\right\}
\end{aligned}
$$

Question: Should $\operatorname{Th}(\{a, b\})$ be an extension?

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Answer: No!
a can only be derived if we already have derived $b$.
$b$ can only be derived if we already have derived $a$.

## Groundedness

## Example

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\begin{aligned}
W & =\emptyset \\
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Answer: No!
$a$ can only be derived if we already have derived $b$. $b$ can only be derived if we already have derived $a$.

## Extensions (formally)

## Definition

Let $\Delta=\langle D, W\rangle$ be a closed default theory.
Let $E$ be any set of closed formulae.
Define:

$$
\begin{aligned}
E_{0} & =W \\
E_{i} & =\operatorname{Th}\left(E_{i-1}\right) \cup\left\{\gamma \left\lvert\, \frac{\alpha: \beta}{\gamma} \in D\right., \alpha \in E_{i-1}, \neg \beta \notin E\right\}
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$$

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$E$ is called an extension of $\Delta$ if

$$
E=\bigcup_{i=0}^{\infty} E_{i}
$$

## How to use this definition?

- The definition does not tell us how to construct an extension.
- However, it tells us how to check whether a set is an extension:
1 Guess a set $E$.
2 Then construct sets $E_{i}$ by starting with $W$.
3 If $E=\bigcup_{i=0}^{\infty} E_{i}$, then $E$ is an extension of $\langle D, W\rangle$.

$$
\begin{array}{ll}
D=\left\{\frac{a: b}{b}, \frac{b: a}{a}\right\} & W=\{a \vee b\} \\
D=\left\{\frac{a: b}{\neg b}\right\} & W=\emptyset \\
D=\left\{\frac{a: b}{\neg b}\right\} & W=\{a\} \\
D=\left\{\frac{: a}{a}, \frac{: b}{b}, \frac{: c}{c}\right\} & W=\{b \rightarrow \neg a \wedge \neg c\} \\
D=\left\{\frac{: c}{\neg d}, \frac{: d}{\neg e}, \frac{e}{\neg f}\right\} & W=\emptyset \\
D=\left\{\frac{: c}{\neg d}, \frac{: d}{\neg c}\right\} & W=\emptyset \\
D=\left\{\frac{a: b}{c}, \frac{a: d}{e}\right\} & W=\{a, \neg b \vee \neg d\}
\end{array}
$$

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## Questions, questions, questions ...

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- What can we say about the existence of extensions?
- How are the different extensions related to each other?
- Can one extension be a subset of another one?
- Are extensions pairwise incompatible (i.e. jointly inconsistent)?
- Can an extension be inconsistent?


## Properties of extensions: existence

## Theorem

11 If $W$ is inconsistent, there is only one extension.
2 A closed default theory $\langle D, W\rangle$ has an inconsistent extensions $E$ if and only if $W$ is inconsistent.

## Proof idea.

If IW' is "inconsistent, no default rule is applicable and $T h(W)$ is the only extension (which is inconsistent as well)

Claim $1 \Longrightarrow$ the if-part.
For only if: Let $W$ be consistent and assume that there exists an
inconsistent extension E
Then there exists a consistent $E_{i}$ such that $E_{i+1}$ is inconsistent.
That is, there is at least one applied default $\alpha_{i}: \beta_{i} / \gamma_{i}$ with

But this contradicts the inconsistency of $E$

## Properties of extensions: existence

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11 If $W$ is inconsistent, there is only one extension.
2 A closed default theory $\langle D, W\rangle$ has an inconsistent extensions $E$ if and only if $W$ is inconsistent.

## Proof idea.

1 If $W$ is inconsistent, no default rule is applicable and $\operatorname{Th}(W)$ is the only extension (which is inconsistent as well).

2 Claim $1 \Longrightarrow$ the if-part.
For only if: Let $W$ be consistent and assume that there exists an inconsistent extension $E$.
Then there exists a consistent $E_{i}$ such that $E_{i+1}$ is inconsistent.
That is, there is at least one applied default $\alpha_{i}: \beta_{i} / \gamma_{i}$ with $\gamma_{i} \in E_{i+1} \backslash \operatorname{Th}\left(E_{i}\right), \alpha_{i} \in E_{i}$, and $\neg \beta_{i} \notin E$.
But this contradicts the inconsistency of $E$.

## Properties of extensions

## Theorem <br> If $E$ and $F$ are extensions of $\langle D, W\rangle$ such that $E \subseteq F$, then $E=F$.

## Proof sketch.

$E=\bigcup E_{i}$ and $F=\bigcup F_{i}$. Use induction to show $F_{i} \subseteq E_{i}$
Base case $i=0$ : Trivially $E_{0}=F_{0}=W$.
Inductive case $i \geq 1$ : Assume $\gamma \in F_{i+1}$. Two cases:
$\gamma \in \operatorname{Th}\left(F_{i}\right)$ implies $\gamma \in \operatorname{Th}\left(E_{i}\right)$ (because $F_{i} \subseteq E_{i}$ by $\| H$ ), and therefore $\gamma \in E_{i+1}$.

Otherwise $\frac{\alpha: \beta}{\gamma} \in D, \alpha \in F_{i}, \neg \beta \notin F$. However, then we have
$\alpha \in E_{i}$ (because $F_{i} \subseteq E_{i}$ ) and $\neg \beta \notin E$ (because of $E \subseteq F$ ), i.e.

## Properties of extensions

## Theorem

If $E$ and $F$ are extensions of $\langle D, W\rangle$ such that $E \subseteq F$, then $E=F$.

## Proof sketch.

$E=\bigcup_{i=0}^{\infty} E_{i}$ and $F=\bigcup_{i=0}^{\infty} F_{i}$. Use induction to show $F_{i} \subseteq E_{i}$.
Base case $i=0$ : Trivially $E_{0}=F_{0}=W$.
Inductive case $i \geq 1$ : Assume $\gamma \in F_{i+1}$. Two cases:
$1 \gamma \in \operatorname{Th}\left(F_{i}\right)$ implies $\gamma \in \operatorname{Th}\left(E_{i}\right)$ (because $F_{i} \subseteq E_{i}$ by IH ), and therefore $\gamma \in E_{i+1}$.

2 Otherwise $\frac{\alpha: \beta}{\gamma} \in D, \alpha \in F_{i}, \neg \beta \notin F$. However, then we have $\alpha \in E_{i}$ (because $F_{i} \subseteq E_{i}$ ) and $\neg \beta \notin E$ (because of $E \subseteq F$ ), i.e., $\gamma \in E_{i+1}$.

## Normal default theories

All defaults in a normal default theory are normal:

$$
\frac{\alpha: \beta}{\beta} .
$$

## Theorem

Normal default theories have at least one extension.
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where $T_{i}$ is a maximal set s.t. (1) $E_{i} \cup T_{i}$ is consistent and (2) if $\beta \in T_{i}$
then there is $\frac{\alpha: \beta}{\rho} \in D$ and $\alpha \in E_{i}$


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If $1 / /$ inconsistent, trivial. Otherwise construct

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where $T_{i}$ is a maximal set s.t. (1) $E_{i} \cup T_{i}$ is consistent and (2) if $\beta \in T_{i}$
$\square$
Show: $T_{i}=\left\{\beta \frac{\alpha: \beta}{\beta} \in D, \alpha \in E_{i}, \neg \beta \notin E\right\}$ for all $i \geq 0$.

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Theorem
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Proof sketch.
If $W$ inconsistent, trivial. Otherwise construct

$$
\begin{aligned}
E_{0} & =W \\
E_{i+1} & =\operatorname{Th}\left(E_{i}\right) \cup T_{i} \quad E=\bigcup_{i=0}^{\infty} E_{i}
\end{aligned}
$$

where $T_{i}$ is a maximal set s.t. (1) $E_{i} \cup T_{i}$ is consistent and (2) if $\beta \in T_{i}$ then there is $\frac{\alpha: \beta}{\beta} \in D$ and $\alpha \in E_{i}$.
Show: $T_{i}=\left\{\beta \left\lvert\, \frac{\alpha: \beta}{\beta} \in D\right., \alpha \in E_{i}, \neg \beta \notin E\right\}$ for all $i \geq 0$.

## Normal default theories: extensions are orthogonal

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## Normal default theories: extensions are orthogonal

## 3世

Proof.
Let $E=\bigcup E_{i}$ and $F=\bigcup F_{i}$ with

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and the same for $F$. Since $E \neq F$, there exists a smallest $i$ such that $E_{i+1} \neq F_{i+1}$.


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## Normal default theories: extensions are orthogonal

## Theorem (Orthogonality)

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and the same for $F$. Since $E \neq F$, there exists a smallest $i$ such that $E_{i+1} \neq F_{i+1}$. This means there exists $\frac{\alpha: \beta}{\beta} \in D$ with $\alpha \in E_{i}=F_{i}$, but with, say, $\beta \in E_{i+1}$ and $\beta \notin F_{i+1}$.

## Normal default theories: extensions are orthogonal

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## Default proofs in normal default theories

## Definition

A default proof of $\gamma$ in a normal default theory $\langle D, W\rangle$ is a finite sequence of defaults $\left(\delta_{i}=\frac{\alpha_{i}: \beta_{i}}{\beta_{i}}\right)_{i=1, \ldots, n}$ in $D$ such that
$1 W \cup\left\{\beta_{1}, \ldots, \beta_{n}\right\} \vdash \gamma$,
$2 W \cup\left\{\beta_{1}, \ldots, \beta_{n}\right\}$ is consistent, and
3 $W \cup\left\{\beta_{1}, \ldots, \beta_{k}\right\} \vdash \alpha_{k+1}$, for $0 \leq k \leq n-1$.

## Theorem

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Let $\Delta=\langle D . W\rangle$ be a normal default theory so that $W$ is consistent. Then $\gamma$ has a default proof in $\Delta$ if and only if there exists an extension $E$ of $\Delta$ such that $\gamma \in E$.

Test 2 (consistency) in the proof procedure suggests that default provability is not even semi-decidable.

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Test 2 (consistency) in the proof procedure suggests that default provability is not even semi-decidable.

## Decidability

## Theorem

Let $\langle D, W\rangle$ be a default theory with $W=\emptyset$ and $D=\left\{\frac{: \beta}{\beta}\right\}$ with $\beta$ an arbitrary closed FOL formula. Clearly, $\beta$ is in some/all extensions of $\langle D, W\rangle$ if and only if $\beta$ is satisfiable.
The existence of a semi-decision procedure for default proofs implies that there is a semi-decision procedure for satisfiability in FOL.
But this is not possible because FOL validity is semi-decidable and this together with semi-decidability of FOL satisfiability would imply decidability of FOL, which is not the case.

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## Propositional default logic

- Propositional DL is decidable.
- How difficult is reasoning in propositional DL?

The skeptical default reasoning problem
(does $\varphi$ follow from $\Delta$ skeptically: $\Delta \mid \sim \varphi$ ?) is called PDS, credulous reasoning is called LPDS.

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PDS is coNP-hard:
consider $D=\emptyset, W=\emptyset$
IPDS is NP-hard:
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- LPDS is NP-hard:
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## Skeptical reasoning in propositional DL

Lemma
$P D S \in \Pi_{2}^{p}$.
Proof sketch.
M/a show that the complementary problem UNPDS (is there anextension $E$ such that $\varphi \notin E)$ is in $\Sigma_{2}^{P}$. The algorithm:

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## Skeptical reasoning in propositional DL

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We show that the complementary problem UNPDS (is there an extension $E$ such that $\varphi \notin E$ ) is in $\Sigma_{2}^{p}$. The algorithm:

1 Guess set $T \subseteq D$ of defaults, those that are applied.
Verify that defaults in $T$ lead to $E$, using a SAT oracle and the
guessed $E:=\operatorname{Th}\left(\left\{\gamma: \frac{\alpha: \beta}{\gamma} \in T\right\} \cup W\right)$
Verify that $\left\{\gamma \cdot \frac{\alpha: \beta}{\gamma} \in T\right\} \| W H \in$ (SAT oracle)
$\rightsquigarrow$ UNPDS $\in \Sigma_{2}^{p}$.

## Similar: LPDS $\in \Sigma_{2}^{p}$.

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1 Guess set $T \subseteq D$ of defaults, those that are applied.
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${ }^{3}$ Verify that $\left\{\gamma: \frac{\alpha: \beta}{\gamma} \in T\right\} \cup W \nvdash \varphi$ (SAT oracle).
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## Similar: LPDS $\in \Sigma_{2}^{p}$.

## Lemma

Default Logic
Proof sketch.
Reduction from 2QBF to UNPDS:

and $\vec{b}=b_{1}, \ldots, b_{m}$ construct $\Delta=\langle D, W\rangle$ with


## No extension contains both $a_{j}$ and $\neg a_{j}$. Then:

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No extension contains both $a_{i}$ and $\neg a_{i}$.

## Lemma

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No extension contains both $a_{i}$ and $\neg a_{i}$. Then:

# iff there is $E$ s.t. $\varphi(\vec{a}, \vec{b}) \in E$ (by 

iff there is $\Delta \subset\{a, \neg a, \ldots, ~ a \rightarrow a\}$ s.t. $A=\varphi(\vec{a}, \vec{b})$ iff $\exists \vec{a} \forall \vec{b} \varphi(\vec{a}, \vec{b})$ is true.

## Lemma

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No extension contains both $a_{i}$ and $\neg a_{i}$. Then:
$\Delta \mid \nsim \neg \varphi(\vec{a}, \vec{b})$ iff there is an extension $E$ s.t. $\neg \varphi(\vec{a}, \vec{b}) \notin E$

## Lemma

PDS is $\Pi_{2}^{p}$-hard.

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No extension contains both $a_{i}$ and $\neg a_{i}$. Then:
iff there is an extension $E$ s.t. $\neg \varphi(\vec{a}, \vec{b}) \notin E$ iff there is $E$ s.t. $\varphi(\vec{a}, \vec{b}) \in E\left(\right.$ by $\left.\frac{: \varphi(\vec{a}, \vec{b})}{\varphi(\vec{a}, \vec{b})} \in D\right)$

## Lemma

PDS is $\Pi_{2}^{p}$-hard.

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PDS is $\Pi_{2}^{p}$-hard.

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## Lemma

PDS is $\Pi_{2}^{p}$-hard.
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## Proof sketch.

Reduction from 2QBF to UNPDS: For $\exists \vec{a} \forall \vec{b} \varphi(\vec{a}, \vec{b})$ with $\vec{a}=a_{1}, \ldots, a_{n}$ and $\vec{b}=b_{1}, \ldots, b_{m}$ construct $\Delta=\langle D, W\rangle$ with

$$
D=\left\{\frac{: a_{i}}{a_{i}}, \frac{: \neg a_{i}}{\neg a_{i}}, \frac{: \varphi(\vec{a}, \vec{b})}{\varphi(\vec{a}, \vec{b})}\right\}, W=\emptyset
$$

No extension contains both $a_{i}$ and $\neg a_{i}$. Then:
$\Delta \mid \nsim \neg \varphi(\vec{a}, \vec{b})$ iff there is an extension $E$ s.t. $\neg \varphi(\vec{a}, \vec{b}) \notin E$
iff there is $E$ s.t. $\varphi(\vec{a}, \vec{b}) \in E\left(\right.$ by $\left.\frac{: \varphi(\vec{a}, \vec{b})}{\varphi(\vec{a}, \vec{b})} \in D\right)$
iff there is $A \subseteq\left\{a_{1}, \neg a_{1}, \ldots, a_{n}, \neg a_{n}\right\}$ s.t. $A \models \varphi(\vec{a}, \vec{b})$
iff $\exists \vec{a} \forall \vec{b} \varphi(\vec{a}, \vec{b})$ is true.

## Conclusions \& remarks

## Theorem

# PDS is $\Pi_{2}^{p}$-complete, even for defaults of the form $\frac{: \alpha}{\alpha}$. 

Theorem
LPDS is $\sum_{2}^{p}$-complete, even for defaults of the form $\frac{\alpha}{\alpha}$.

> PDS is "easier" than reasoning in most modal logics
> General and normal defaults have the same complexity.
> Polynomial special cases cannot be achieved by restricting
> for example, to Horn clauses (satisfiability testing in
> polynomial time)
> It is necessary to restrict the underlying monotonic reasoning problem and the number of extensions

> Similar results hold for other nonmonotonic logics.

## Conclusions \& remarks

## Theorem

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## Theorem

- PDS is "easier" than reasoning in most modal logics.
- General and normal defaults have the same complexity.
- Polynomial special cases cannot be achieved by restricting, for example, to Horn clauses (satisfiability testing in polynomial time).
- It is necessary to restrict the underlying monotonic reasoning problem and the number of extensions.
- Similar results hold for other nonmonotonic logics.

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## Semi-normal defaults (1)

Semi-normal defaults are sometimes useful:


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For Student (TOM) we get two extensions: one with
Employed (TOM) and the other one with $\neg$ Employed (Tom).
Since the third rule is "more specific", we may prefer it.

## Semi-normal defaults (1)

Semi-normal defaults are sometimes useful:

$$
\frac{\alpha: \beta \wedge \gamma}{\beta}
$$

Important when one has interacting defaults:
$\frac{\text { Adult }(x) \text { : Employed }(x)}{\text { Employed }(x)}$
$\frac{\text { Student }(x): \text { Adult }(x)}{\text { Adult }(x)}$
$\frac{\text { Student }(x): \neg \text { Employed }(x)}{\neg \operatorname{Employed}(x)}$

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$$
\begin{gathered}
\frac{\text { Adult }(x): \text { Employed }(x)}{\text { Employed }(x)} \\
\frac{\text { Student }(x): \operatorname{Adult}(x)}{\text { Adult }(x)} \\
\frac{\text { Student }(x): \neg \operatorname{Employed}(x)}{\neg \operatorname{Employed}(x)}
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\end{aligned}
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For Student (TOM) we get two extensions: one with Employed (TOM) and the other one with $\neg$ Employed(Tom). Since the third rule is "more specific", we may prefer it.

## Semi-normal defaults (2)

- Since being a student is an exception, we could use a semi-normal default to exclude students from employed adults:


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Student $(x)$ : Adult $(x)$
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> Representing conflict-resolution by semi-normal defaults becomes clumsy when the number of default rules
> becomes high.
> A scheme for assigning priorities would be more elegant (there are indeed such schemes).

## Semi-normal defaults (2)

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$$
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Representing conflict-resolution by semi-normal defaults becomes clumsy when the number of default rules
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Semi-normal defaults

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- Representing conflict-resolution by semi-normal defaults becomes clumsy when the number of default rules becomes high.
- A scheme for assigning priorities would be more elegant (there are indeed such schemes).


## Open defaults (1)

- Our examples included open defaults, but the theory covers only closed defaults.
- If we have $\frac{\alpha(\vec{x}): \beta(\vec{x})}{\gamma(\vec{x})}$, then the variables should stand for all nameable objects.
- Problem: What about objects that have been introduced implicitly, e.g., via formulae such a $\exists x P(x)$.
- Solution by Reiter: Skolemization of all formulae in $W$ and D.
- Interpretation: An open default stands for all the closed defaults resulting from substituting ground terms for the variables.


## Open defaults (2)

## Skolemization can create problems because it preserves satisfiability, but it is not an equivalence transformation

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Skolemization of $\exists y$ : ... enables concluding Bachelor(TOM)! The reason is that for $a(T O M)$ we det $\operatorname{Man}(a(T O M))$ bv default (where $g$ is the Skolem function).

## Open defaults (2)

Skolemization can create problems because it preserves satisfiability, but it is not an equivalence transformation.

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```
\forallx(\operatorname{Man}(x)\leftrightarrow\negWoman}(x)
\forall(Man(x) ->(\existsy(Spouse(x,y)^Woman}(y))\vee\operatorname{Bachelor}(x))
Man(TOM)
Spouse(TOM, MARY)
Woman(MARY)
\[
\frac{: \operatorname{Man}(x)}{\operatorname{Man}(x)}
\]
```

Skolemization of $\exists y$ : ... enables concluding Bachelor(TOM)! The reason is that for $g(T O M)$ we get $\operatorname{Man}(g(T O M))$ by default (where $g$ is the Skolem function).

## Open defaults (3)

It is even worse: Logically equivalent theories can have different extensions:

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$W_{1}$ and $W_{2}$ are logically equivalent. However, the Skolemization of $W_{1}$, symbolically $s\left(W_{1}\right)$, is not equivalent with $s\left(W_{2}\right)$. The only extension of $\left\langle D, W_{1}\right\rangle$ is $\operatorname{Th}\left(s\left(W_{1}\right) \cup R\right)$. The only extension of $\left\langle D, W_{2}\right\rangle$ is $\operatorname{Th}\left(s\left(W_{2}\right)\right)$.
Note: Skolemization is not the right method to deal with open defaults in the general case.

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It is even worse: Logically equivalent theories can have different extensions:

$$
\begin{aligned}
W_{1} & =\{\exists x(P(c, x) \vee Q(c, x))\} \\
W_{2} & =\{\exists x P(c, x) \vee \exists x Q(c, x)\} \\
D & =\left\{\frac{P(x, y) \vee Q(x, y): R}{R}\right\}
\end{aligned}
$$

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\end{aligned}
$$

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$W_{1}$ and $W_{2}$ are logically equivalent. However, the Skolemization of $W_{1}$, symbolically $s\left(W_{1}\right)$, is not equivalent with $s\left(W_{2}\right)$. The only extension of $\left\langle D, W_{1}\right\rangle$ is $\operatorname{Th}\left(s\left(W_{1}\right) \cup R\right)$. The only extension of $\left\langle D, W_{2}\right\rangle$ is $\operatorname{Th}\left(s\left(W_{2}\right)\right)$.

Note: Skolemization is not the right method to deal with open defaults in the general case

It is even worse: Logically equivalent theories can have different extensions:

$$
\begin{aligned}
W_{1} & =\{\exists x(P(c, x) \vee Q(c, x))\} \\
W_{2} & =\{\exists x P(c, x) \vee \exists x Q(c, x)\} \\
D & =\left\{\frac{P(x, y) \vee Q(x, y): R}{R}\right\}
\end{aligned}
$$

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## Outlook

Although Reiter's definition of DL makes sense, one can come model-theoretically).

- General proof methods for the finite, decidable case
- Applications of default logic:
- Diagnosis
- Reasoning about actions


## Literature

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