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## Non-Monotonic Reasoning II

Lecture 8, 24th Nov 2024 // Foundations of Knowledge Representation, WS 2025/26

## **Datalog & Least Herbrand Models**

#### We have seen so far:

- It is easy to formalise intuitions about preferred models if we have a least Herbrand model.
- In that case, everyone agrees that the least Herbrand model is the right choice
- Datalog knowledge bases have a least Herbrand model, which can be computed deterministically using forward chaining
- We can successfully formalise the Closed World Assumption

However, we cannot express default statements:

 $\frac{\textit{hasOrg}(x,y) \land \textit{Heart}(y) \& \text{ consistent to assume } \textit{hasLocation}(y,\textit{left})}{\text{deduce } \textit{hasLocation}(y,\textit{left})}$ 





# **Going Beyond Datalog**

To overcome expressivity limitations we next

- 1. Extend Datalog to a more expressive logic
- 2. Develop a new mechanism for selecting preferred models

Idea: First, allow for negation in the body of rules:

$$\forall x \forall y \Big( (hasOrg(x,y) \land Heart(y) \land \sim hasLocation(y, right)) \rightarrow hasLocation(y, left) \Big)$$

Then, devise a preferred model selection mechanism such that negation is read non-monotonically, as follows:

- "Deduce that heart is on the left unless we can deduce that it is on the right"
- "Deduce that the heart is on the left if ~hasLocation(y, right) (that is, hasLocation(y, left)) is consistent with our knowledge"





# Datalog¬-Rules

A Datalog¬ rule is a function-free, universally quantified implication of the form

$$(L_1 \wedge \ldots \wedge L_n) \rightarrow H$$

with  $L_i$  a literal (an atom A or negated atom  $\sim A$ ) and H either an atom or  $\bot$ . A Datalog $^\neg$  knowledge base is a pair  $\mathcal{K} = \langle \mathcal{R}, \mathcal{F} \rangle$  where  $\mathcal{R}$  is a finite set of Datalog $^\neg$  rules and  $\mathcal{F}$  is a finite set of facts.

$$\forall x.(Heart(x) \land hasLoc(x, left) \rightarrow SitSolHeart(x))$$

$$\forall x.(Heart(x) \land hasLoc(x, right) \rightarrow SitInvHeart(x))$$

$$\forall x. \forall y. (Human(x) \land hasOrg(x, y) \land SitInvHeart(y) \rightarrow SitInvPatient(x))$$

$$\forall x. \forall y. (Human(x) \land hasOrg(x, y) \land SitSolHeart(y) \rightarrow Healthy(x))$$

$$\forall x. (\forall y. (hasOrg(x, y) \land Heart(y) \land \sim hasLoc(y, right) \rightarrow hasLoc(y, left)))$$





So far all this is just syntax.

We need to specify the semantics of Datalog¬.

→ Which are the preferred models?

There was a "war of semantics" in 1980s and 1990s.

Meaning of things like  $\sim B \rightarrow A$  and  $\sim A \rightarrow B$ 

Single vs. multiple-models semantics

To date, we have the following:

- Well-founded Semantics
- Stable Model Semantics (aka Answer Set Semantics)





We will focus on Stable Model Semantics.

Preferred models are called Stable Models (SM).

It thus follows that

$$\mathfrak{K} \bowtie a$$
 iff  $\mathfrak{I} \models a$  for each stable model  $\mathfrak{I}$  of  $\mathfrak{K}$ 

We will see that  $\mathfrak K$  may have

- no stable models, or
- one stable model, or
- several stable models.

Furthermore, if  $\mathcal K$  contains only Datalog rules (i.e., no negation), then  $\mathcal K$  has exactly one stable model (the least Herbrand model).





We proceed as follows:

- 1. Define stable models for the propositional case.
- 2. Extend to the case with variables using grounding.

A simple propositional example  $\mathcal K$  with one rule and one fact:

Suspect 
$$\land \sim Guilty \rightarrow Innocent$$
  
Suspect

Intuitively, the rule says the following:

"A suspect is innocent unless they can be proved guilty."

We only know that Suspect holds, so we intuitively expect that

 $\mathcal{K} \approx Innocent$ 





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"A suspect is innocent unless they can be proved guilty."

We only know that Suspect holds, so we intuitively expect that

 $\mathcal{K} \approx Innocent$ 

and  $\mathcal{K} \cup \{Guilty\} \not \models Innocent$ 





Our example: Suspect  $\land \sim Guilty \rightarrow Innocent$ Suspect

Intuitively, the following (Herbrand-style) model should be stable:

$$J_1 = \{Suspect, Innocent\}$$

To check this, we first compute the reduct  $\mathcal{K}^{\mathcal{I}_1}$  of  $\mathcal{K}$  by  $\mathcal{I}_1$ :

- 1. Remove all rules with negative body literal  $\sim A$  such that the (positive) literal A is in  $\mathfrak{I}_1$
- 2. Remove all negative literals from remaining rules

The result is always a Datalog knowledge base.

In our example, we do not remove any rule since *Guilty*  $\notin \mathcal{I}_1$ :

Suspect → Innocent Suspect





Once we have the reduct  $\mathfrak{K}^{\mathfrak{I}_1}$ 

We check whether  $\mathfrak{I}_1$  is the least Herbrand Model of  $\mathfrak{K}^{\mathfrak{I}_1}$ , in which case  $\mathfrak{I}_1$  is a stable model.

Indeed, by using forward chaining we can see that

$$J_1 = \{Suspect, Innocent\}$$

is the least Herbrand model of  $\mathcal{K}^{\mathcal{I}_1}$  and hence  $\mathcal{I}_1$  is a stable model of  $\mathcal{K}$ .

But this is not sufficient to show  $\mathfrak{K} \approx Innocent$ .

 $\rightsquigarrow$  We need to look at all stable models of  $\mathfrak{K}$ .





Let us check the remaining possibilities:

```
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```

The reducts  $\mathcal{K}^{J_2}$  and  $\mathcal{K}^{J_3}$  are the same and contain just the fact:

#### Suspect

This is so because  $Guilty \in \mathcal{I}_2, \mathcal{I}_3$  and hence the reduct does not include the only rule we have in  $\mathcal{K}$ .

The least model of  $\mathcal{K}^{\mathcal{I}_2}$  (or  $\mathcal{K}^{\mathcal{I}_3}$ ) is  $\mathcal{I}_4$ , thus neither  $\mathcal{I}_2$  nor  $\mathcal{I}_3$  are stable.





We finally check whether

$$\mathfrak{I}_4 = \{Suspect\}$$

is a stable model of

Suspect 
$$\land \sim Guilty \rightarrow Innocent$$
  
Suspect

The reduct  $\mathcal{K}^{\mathfrak{I}_4}$  is the same as  $\mathcal{K}^{\mathfrak{I}_1}$ , namely

But then  $\mathfrak{I}_4$  is not even a model of  $\mathfrak{K}^{\mathfrak{I}_4}$ .

Thus,  $\mathfrak{I}_1 = \{Suspect, Innocent\}$  is the only stable model of  $\mathfrak{K}$  and so

 $\mathfrak{K} \approx Innocent.$ 





Consider K as follows:

```
~Guilty → Innocent
~Innocent → Guilty
```

Recall that we compute the reduct  $\mathcal{K}^{\mathcal{I}}$  of  $\mathcal{K}$  by  $\mathcal{I}$  as follows:

- 1. Remove all rules with negative body literal  $\sim\!\!A$  such that the (positive) literal A is in  $\mathcal I$
- 2. Remove all negative literals from remaining rules

SM candidates: Ø, {Guilty}, {Innocent}, {Guilty, Innocent}





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- 2. Remove all negative literals from remaining rules

```
SM candidates: \emptyset, \{Guilty\}, \{Innocent\}, \{Guilty, Innocent\}
Reduct \mathbf{K}^{\emptyset}:

Innocent

Guilty
```

Least model of  $\mathbf{K}^{\emptyset}$ : {Guilty, Innocent}  $\neq \emptyset \rightsquigarrow$  no stable model





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SM candidates: \emptyset, \{Guilty\}, \{Innocent\}, \{Guilty, Innocent\}
Reduct \mathbf{K}^{\{Guilty\}}:
```

Least model of  $\mathbf{K}^{\{Guilty\}}$ :  $\{Guilty\} \rightsquigarrow \text{stable model}$ 





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```
SM candidates: \emptyset, {Guilty}, {Innocent}, {Guilty, Innocent}
Reduct \mathbf{K}^{\{Innocent\}}:
```

Innocent

Least model of **K**<sup>{Innocent</sup>}: {Innocent} → stable model





Consider K as follows:

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~Innocent → Guilty
```

Recall that we compute the reduct  $\mathcal{K}^{J}$  of  $\mathcal{K}$  by  $\mathcal{I}$  as follows:

- 1. Remove all rules with negative body literal  $\sim\!\!A$  such that the (positive) literal A is in  $\Im$
- 2. Remove all negative literals from remaining rules

```
SM candidates: \emptyset, {Guilty}, {Innocent}, {Guilty, Innocent}
Reduct \mathbf{K}^{\{Guilty, Innocent\}}: (empty set)
```

Least model of  $\mathbf{K}^{\{Guilty, Innocent\}}$ :  $\emptyset \neq \{Guilty, Innocent\} \rightsquigarrow \text{no stable model}$ 





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Stable models: {Guilty}, {Innocent}





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SM candidates: Ø, {Guilty}, {Innocent}, {Guilty, Innocent}
```

Stable models: {Guilty}, {Innocent}

→ A KB can have several stable models.





Consider  $\mathfrak K$  as follows:

$$\sim$$
Guilty  $\rightarrow$  Guilty

Recall that we compute the reduct  $\mathcal{K}^{\mathbb{J}}$  of  $\mathcal{K}$  by  $\mathbb{J}$  as follows:

- 1. Remove all rules with negative body literal  $\sim\!\!A$  such that the (positive) literal A is in  $\Im$
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Stable model candidates: Ø, {*Guilty*}





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Stable model candidates: Ø, {*Guilty*}

Reduct **K**<sup>Ø</sup>:

Guilty

Least model of  $\mathbf{K}^{\emptyset}$ : {*Guilty*}  $\neq \emptyset \rightsquigarrow$  no stable model





Consider  $\mathfrak K$  as follows:

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```
Stable model candidates: Ø, {Guilty}
```

Reduct  $\mathbf{K}^{\{Guilty\}}$ : (empty set)

Least model of  $\mathbf{K}^{\{Guilty\}}$ :  $\emptyset \neq \{Guilty\} \rightsquigarrow$  no stable model





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Stable model candidates: Ø, {*Guilty*}

→ A KB may have no stable models.





# Non-monotonic vs. Classical Negation

Consider again our propositional example  $\mathfrak{K}$ :

Suspect 
$$\land \neg Guilty \rightarrow Innocent$$
  
Suspect

Let us check whether

$$\mathcal{K} \models Innocent$$

for  $\models$  being entailment under monotonic PL semantics.

Clearly, K is equivalent in standard propositional logic to

Hence  $\mathcal{I} = \{Suspect, Guilty\}$  is a model of  $\mathcal{K}$  with  $\mathcal{I} \not\models Innocent$ , thus:

 $\mathcal{K} \not\models Innocent$ 





# **Properties**

Let  $\mathcal{K}$  be a (propositional) Datalog knowledge base. Then:

#### Theorem

Every stable model of  $\mathfrak K$  is a classical model of  $\mathfrak K$ .

#### Corollary

If  $\mathfrak{K} \models a$ , then  $\mathfrak{K} \models a$ .

#### Theorem

If a proposition P holds in some stable model of  $\mathcal{K}$ , then P is a head (or a fact) of some rule in  $\mathcal{K}$ .

#### Theorem

If  $\mathcal{I}_1$  and  $\mathcal{I}_2$  are stable models of  $\mathcal{K}$ , then neither  $\mathcal{I}_1 \subsetneq \mathcal{I}_2$  nor  $\mathcal{I}_2 \subsetneq \mathcal{I}_1$ .





So far, all this is propositional. What about ...

```
\forall x (Heart(x) \land hasLoc(x, left) \rightarrow SitSolHeart(x)) \\ \forall x (Heart(x) \land hasLoc(x, right) \rightarrow SitInvHeart(x)) \\ \forall x \forall y (Human(x) \land hasOrg(x, y) \land SitInvHeart(y) \rightarrow SitInvPatient(x)) \\ \forall x \forall y (Human(x) \land hasOrg(x, y) \land SitSolHeart(y) \rightarrow Healthy(x)) \\ \forall x (\forall y (hasOrg(x, y) \land Heart(y) \land \sim hasLoc(y, right) \rightarrow hasLoc(y, left))) \\ Human(MJ) \\ hasOrg(MJ, h) \\ Heart(h)
```

Fortunately, we are still within the Bernays-Schönfinkel class.\* We can apply grounding and reduce to the propositional case.

<sup>\*:</sup> Bernays-Schönfinkel formulas are of the form  $\exists x_1 \dots \exists x_m \forall y_1 \dots \forall y_n \varphi$  with  $\varphi$  quantifier-free.





So, to compute all the stable models of  $\mathcal{K}$ :

- 1. Compute the grounding of  ${\mathfrak K}$  over the Herbrand universe.
- 2. Compute all the stable models of the resulting propositional KB. Obviously, the grounding could be of exponential size.

But this is a computational hazard, not a conceptual one.





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But this is a computational hazard, not a conceptual one.

Intuitively, the following Herbrand model should be stable:

```
J_1 = \{Human(MJ), hasOrg(MJ, h), Heart(h), hasLoc(h, left), SitSolHeart(h), Healthy(MJ)\}
```





So, to compute all the stable models of  $\mathfrak{K}$ :

- 1. Compute the grounding of  ${\mathfrak K}$  over the Herbrand universe.
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But this is a computational hazard, not a conceptual one.

Intuitively, the following Herbrand model should be stable:

```
J_1 = \{Human(MJ), hasOrg(MJ, h), Heart(h), hasLoc(h, left), SitSolHeart(h), Healthy(MJ)\}
```

On the other hand, the following one should not be stable:

```
J_2 = \{Human(MJ), hasOrg(MJ, h), Heart(h), hasLoc(h, right), SitInvHeart(h), SitInvPatient(MJ)\}
```





To check whether

```
J_1 = \{Human(MJ), hasOrg(MJ, h), Heart(h), hasLoc(h, left), SitSolHeart(h), Healthy(MJ)\}
```

is stable, notice that even though the grounding is huge, the only PL formulas that matter are the following:

```
Heart(h) \land hasLoc(h, left) \rightarrow SitSolHeart(h)

Human(MJ) \land hasOrg(MJ, h) \land SitSolHeart(h) \rightarrow Healthy(MJ)

hasOrg(MJ, h) \land Heart(h) \land \sim hasLoc(h, right) \rightarrow hasLoc(h, left)

Human(MJ), hasOrg(MJ, h), Heart(h)
```

The reduct of  $I_1$  over those formulas is

```
Heart(h) \land hasLoc(h, left) \rightarrow SitSolHeart(h)

Human(MJ) \land hasOrg(MJ, h) \land SitSolHeart(h) \rightarrow Healthy(MJ)

hasOrg(MJ, h) \land Heart(h) \rightarrow hasLoc(h, left)

Human(MJ), hasOrg(MJ, h), Heart(h)
```

And clearly,  $\mathcal{I}_1$  is the least model.





To check whether

```
I_2 = \{Human(MJ), hasOrg(MJ, h), Heart(h), hasLoc(h, right), SitInvHeart(h), SitInvPatient(MJ)\}
```

is stable, the relevant PL formulas are the following:

```
Heart(h) \land hasLoc(h, right) \rightarrow SitInvHeart(h)

Human(MJ) \land hasOrg(MJ, h) \land SitInvHeart(h) \rightarrow SitInvPatient(MJ)

hasOrg(MJ, h) \land Heart(h) \land \sim hasLoc(h, right) \rightarrow hasLoc(h, left)

Human(MJ), hasOrg(MJ, h), Heart(h)
```

The reduct of  $I_2$  over those formulas is

```
Heart(h) \land hasLoc(h, right) \rightarrow SitInvHeart(h)
Human(MJ) \land hasOrg(MJ, h) \land SitInvHeart(h) \rightarrow SitInvPatient(MJ)
Human(MJ), hasOrg(MJ, h), Heart(h)
```

And clearly,  $\mathcal{I}_2$  is not the least model.





# **Quick Recap**

We have seen that by using Datalog with non-monotonic negation

- 1. We can formalise the closed-world assumption
- 2. We can express default statements

The key notion is that of a Stable Model as a "preferred" model.

Checking whether a propositional model is stable involves

- 1. Eliminating negation by computing the reduct
- 2. Checking if the candidate is the least model of the reduct

Checking whether a FOL Herbrand interpretation is a stable model involves

- 1. Computing the propositional grounding of the KB
- 2. Checking whether the candidate is stable for the grounding

Note: Stable models in the FOL case are always Herbrand models.





#### What have we left out?

Much more than we have covered!

The field of NMR is huge and we have just seen the tip of the iceberg.

Extensions related to what we have seen:

Stable models and disjunctive rules (disjunction in the head), e.g.

$$professor(x), semester(s) \rightarrow teaches(x, s) \lor sabbatical(x, s)$$

- Stable models and general propositional formulas
- Combinations of classical and non-monotonic negation, e.g.

$$suspect(x)$$
,  $\sim guilty(x) \rightarrow \neg guilty(x)$ 





# Relationships with other areas

What we have seen is not only relevant to KR.

There are strong connections with other fields:

- Answer Set Programming (ASP)
   Using negation we can encode search problems
- Deductive databases
  - Database systems that can conclude new data using rules
- Logic programming (Prolog)
  - Negation as failure can help write shorter programs



