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Example

$$KB = \left\{ egin{array}{ll} ext{false :- a,b.} \ ext{a :- c.} \ ext{b :- c.} \end{array} 
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## Satisfiability

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Horn-SAT is feasible, whereas 3-SAT is likely not.

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Example: Assume a database of video segments is complete.

## Rules

Encode birds fly

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to allow for exceptions.

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j is true in some model of KB

```
\frac{\text{bird}(X) : \text{fly}(X)}{\text{fly}(X)}
```

```
Let KB be
   bird(robin).
   bird(penguin).
   false :- fly(penguin).
   fly(bee).
```

#### Conclude:

```
(*) \frac{\text{bird}(X) : \text{fly}(X)}{\text{fly}(X)}
Let KB be
     bird(robin).
     bird(penguin).
     false :- fly(penguin).
     fly(bee).
Conclude:
     fly(robin) by default rule (\star)
but not fly(penguin).
```

### Non-determinism

Conflicting defaults

$$\frac{\operatorname{quaker}(X):\operatorname{pacifist}(X)}{\operatorname{pacifist}(X)} \qquad \frac{\operatorname{republican}(X):\operatorname{hawk}(X)}{\operatorname{hawk}(X)}$$

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Let KB be
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Applying one default to Nixon makes the other inapplicable.

KB has two incompatible extensions, breaking least fixed point (provability model) for Horn clauses.

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N.B. Checking finite failure can be as hard as the Halting Problem.

# 3 modes of inference (C.S. Peirce)

		typed functional prog $\cong$ proof
Deduction	deduce	modus ponens $\cong$ function app $f(a)$
Abduction	explain	choose input a from assumables
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From  $\models$  as inclusion  $\subseteq$ 

$$KB \models g \iff Mod(KB) \subseteq Mod(g)$$
 $KB \text{ satisfiable} \iff Mod(KB) \not\subseteq Mod(false)$ 
 $\iff Mod(KB) \neq \emptyset$ 

to weighing alternatives  $d \in D$  via probabilities given KB

$$prob(d|KB) = conditional probability of d given KB$$

→ Bayesian networks . . .