

# *Artificial Intelligence*

## Plausible Reasoning

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# Beyond classical logic?

- For *classical* logic it usually intended:

First-order logic  $L_{FO}$

Propositional logic  $L_P$  (which is contained in  $L_{FO}$ )

- A **non-classical** logic adopts different rules

- *What for?*

Representing other forms of reasoning

Not just *deduction* but also *abduction* and *induction* (*see after*)

Specialized reasoning, e.g. about time or other modalities like belief, intentions etc.

For practical applications

Subsets of  $L_{FO}$ , that are either more efficient or focused on a specific purpose (e.g. Prolog)

# Logics and logical systems

Theoretically, a **logic** includes:

- a) Formal language
- b) Formal semantics of the language
- c) Relations  $\models$  (*entailment*) e  $\vdash$  (*derivation*)

## ■ In the realm of artificial intelligence

A **logical system** is a *reasoning agent* (not necessarily human)

- It is based on the a **logic** of reference (e.g.  $L_{FO}$ )
- It makes use of a **computation strategy** (e.g. *SLD depth-first*)
- It may have *limited resources* (e.g. time or memory or both)

This leads to the idea of derivability in a logical system

Notation:  $\Gamma \vdash_{\langle SysLog \rangle} \varphi$  where  $\langle SysLog \rangle$  describes a particular logical system

Example:

$$\Gamma \vdash_{L_{FO}} \varphi \neq \Gamma \vdash_{\overset{\text{SLD strategy (just for Horn clauses) that is also fair}}{SLD\ fair}} \varphi \neq \Gamma \vdash_{\overset{\text{A generic SLD strategy (i.e. not necessarily fair)}}{SLD}} \varphi$$

$\uparrow$  General derivability in  $L_{FO}$                        $\nwarrow$

In the line of principle, the computation strategy of  $\langle SysLog \rangle$  can be anything:  
e.g.  $\Gamma \vdash_{NN} \varphi$  might be a neural network that says whether  $\varphi$  is (NN)-derivable from  $\Gamma$

# Defeasible reasoning

A reasoning process where the **relation** between formulae is rationally plausible yet not necessarily correct (in the classical logical sense)

Notation:

$\Gamma \vdash_{\langle \text{SysLog} \rangle} \varphi$  says that  $\varphi$  is a **plausible** derivation from  $\Gamma$  in  $\langle \text{SysLog} \rangle$

Properties of  $\vdash_{\langle \text{SysLog} \rangle}$

$$\Gamma \vdash_{\langle \text{SysLog} \rangle} \varphi \Rightarrow \Gamma \not\vdash_{\langle \text{SysLog} \rangle} \neg \varphi$$

(coherence)

$$\Gamma \vdash_{\langle \text{SysLog} \rangle} \varphi \Rightarrow \Gamma \vdash_{\langle \text{SysLog} \rangle} \varphi$$

(compatibility with derivation)

$$\Gamma \vdash_{\langle \text{SysLog} \rangle} \varphi \not\Rightarrow \Gamma \vdash_{\langle \text{SysLog} \rangle} \varphi (\Rightarrow \Gamma \models \varphi)$$

(not necessarily correct)

Occurs very often in practice:

*“The train schedule does not report a train to Milano at 06:55,  
therefore we assume that such a train does not exist”*

Most databases contain positive information only

Negative facts are often derived ‘by default’

# Defeasible reasoning

- Inference in *defeasible reasoning* is

## Non-monotonic

$$\Gamma \vdash_{\langle \text{SysLog} \rangle} \varphi \not\Rightarrow \Gamma \cup \Delta \vdash_{\langle \text{SysLog} \rangle} \varphi$$

The arrival of new pieces of information may falsify inferences that used to be justified  
*e.g. an extra train to Milano at 06:55 is announced ...*

## Systemic

In classical logic, the soundness of all inferences schema depend only on the few formulae involved.

e.g.  $\varphi \rightarrow \psi, \varphi \vdash \psi$

In defeasible reasoning, inferences are justified by an entire theory  $\Gamma$

One must check the entire database:  $\Gamma \not\vdash \varphi \vdash_{\langle \text{SysLog} \rangle} \neg \varphi$

# Closed-World Assumption (CWA)

$$\{\Gamma \not\models \alpha\} \vdash_{CWA} \neg\alpha \quad (\alpha \text{ is an atom})$$

Example (a program):

$$\Pi \equiv \{\{Philosopher(socrates)\}, \{Philosopher(plato)\}, \{Cat(felix)\}\}$$

The program  $\Pi$  can be rewritten in  $L_{FO}$  as:

$$\forall x ((x = socrates) \rightarrow Philosopher(x))$$

$$\forall x ((x = plato) \rightarrow Philosopher(x))$$

$$\forall x ((x = felix) \rightarrow Cat(x))$$

The *Closed-World Assumption* (CWA) means completing (i.e. extending) the program  $\Pi$ :

$$\forall x ((x = felix) \leftrightarrow Cat(x))$$

$$\forall x ((x = socrates \vee x = plato) \leftrightarrow Philosopher(x)) \quad \text{Notice the double implication}$$

Then these plausible inferences become sound:

$$\Pi \vdash_{CWA} \neg Cat(socrates)$$

$$\Pi \vdash_{CWA} \neg Cat(plato)$$

$$\Pi \vdash_{CWA} \neg Philosopher(felix)$$

# Inference and reasoning (according to C. S. Peirce, 1870 c.a. )

## ■ Different types of reasoning

### Deductive inference (*sound*)

*Derive only what is justified in terms of **entailment***

"All beans in this bag are white"

"This handful of beans comes from the bag"

"This is a handful of white beans"

$$\frac{\begin{array}{l} \forall x \varphi(x) \rightarrow \psi(x) \\ \varphi(a) \end{array}}{\psi(a)}$$

### Inductive inference (*plausible*)

*From repeated occurrences, derive rules*

"This handful of beans comes from the bag"

"This is a handful of white beans"

"All beans in this bag are white"

$$\frac{\begin{array}{l} \psi(a) \\ \varphi(a) \end{array}}{\forall x \varphi(x) \rightarrow \psi(x)}$$

### Abductive inference (*plausible*)

*From rules and outcomes, derive premises*

"All beans in this bag are white"

"This is a handful of white beans"

"This handful of beans comes from the bag"

$$\frac{\begin{array}{l} \forall x \varphi(x) \rightarrow \psi(x) \\ \psi(a) \end{array}}{\varphi(a)}$$

# Abductive inferences: explanatory hypotheses

*The basic theory is still that of classical logic*

What changes is the way reasoning (and hence calculus) is performed

## ■ Abductive reasoning, in general:

A **model** (or abstract definition of some kind)  
represented by a logical theory  $K$

A set of specific **observations**  
represented by a set of wffs  $\Sigma$

In general:  $K \not\models \Sigma$

(specific observations are not *entailed* by the model)

The problem is finding *hypotheses*  $\Delta$  (i.e sets of wffs) such that

$$K \cup \Delta \models \Sigma$$

Intuitively, a set  $\Delta$  describes an hypothesis that *explains* the observations  $\Sigma$



# Example: "The car does not start"

## ■ Model (K)

$\kappa_1: dischargedBattery \rightarrow (\neg lightsOn \wedge \neg radioOn \wedge \neg selfStarterRuns)$

$\kappa_2: selfStarterBroken \rightarrow \neg selfStarterRuns$

$\kappa_3: \neg selfStarterRuns \rightarrow \neg engineStarts$

$\kappa_4: voidTank \rightarrow (gasGaugeZero \wedge \neg engineStarts)$

## ■ Observation ( $\Sigma$ )

$\sigma_1: \neg engineStarts$

## ■ Plausible causes ( $\Delta$ )

$\delta_1: dischargedBattery$  since:  $\{\kappa_1, \kappa_3\} \cup \{\delta_1\} \models \sigma_1$

$\delta_2: selfStarterBroken$  since:  $\{\kappa_2, \kappa_3\} \cup \{\delta_2\} \models \sigma_1$

$\delta_3: voidTank$  since:  $\{\kappa_4\} \cup \{\delta_3\} \models \sigma_1$

# Rationality of hypotheses

- *Plausible*

$K \cup \Delta \cup \Sigma$  must be *satisfiable*

- *Minimal*

There must not be a subset  $\Delta^* \subset \Delta$  such that  $K \cup \Delta^* \models \Sigma$

- *Relevant*

$K \cup \{\neg engineStarts\} \models \neg engineStarts$

is both plausible and minimal but offers no explication

(abductive reasoning is about the *causes*, in some sense)