# Argumentation-based Causal and Counterfactual Reasoning

Lars Bengel<sup>1</sup>, Lydia Blümel<sup>1</sup>, Tjitze Rienstra<sup>2</sup>, Matthias Thimm<sup>1</sup>

 Artificial Intelligence Group, University of Hagen, Germany
 Department of Advanced Computing Sciences, Maastricht University, The Netherlands

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# Argumentation

#### **Definition**

An argumentation framework is a pair  $F = (\mathbf{A}, \Rightarrow)$  where  $\mathbf{A}$  is a set whose elements are called *arguments* and where  $\Rightarrow \subseteq \mathbf{A} \times \mathbf{A}$  is called the *attack relation*.

#### **Definition**

A set  $E \subseteq \mathbf{A}$  is:

- ▶ conflict-free if for all  $a, b \in E$  we have  $a \not\Rightarrow b$ .
- ▶ stable if E is conflict-free and for every  $a \in \mathbf{A} \setminus E$  there is a  $b \in E$  such that  $b \Rightarrow a$ .

Phan Minh Dung. "On the Acceptability of Arguments and its Fundamental Role in Nonmonotonic Reasoning, Logic Programming and n-Person Games". In: Artificial Intelligence 77.2 (1995), pp. 321–358

# Causal Knowledge Bases

#### **Definition**

A causal model is a triple (U, V, K) where U and V partition the set of atoms into, respectively, a set of background and explainable atoms. K consists of a set of Boolean structural equations, one for each atom  $v \in V$ . A Boolean structural equation for v is a formula of the form  $v \leftrightarrow \phi$ .

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#### **Definition**

A causal knowledge base is a knowledge base  $\Delta = (K, A)$  where K is a causal model and where A is a set of background assumptions, at least one for each background atom. A background assumption for an atom u is a literal  $I \in \{u, \neg u\}$ .

# **KB-Induced Argumentation Frameworks**

#### **Definition**

Let  $\Delta = (K, A)$  be a knowledge base. We define the AF induced by  $\Delta$   $F(\Delta)$  as the AF  $(\mathbf{A}, \Rightarrow)$  where

- **A** is the set of all pairs  $(\Phi, \psi)$  such that
  - $ightharpoonup \Phi \subseteq A$ ,
  - $\triangleright \Phi \cup K \nvdash \bot$
  - $\blacktriangleright$   $\Phi \cup K \vdash \psi$ , and if  $\Psi \subset \Phi$  then  $\Psi \cup K \nvdash \psi$ .
- ▶  $\Rightarrow \subseteq \mathbf{A} \times \mathbf{A}$  such that  $((\Phi, \psi), (\Phi', \psi')) \in \Rightarrow$  iff  $(\Phi, \psi)$  undercuts  $(\Phi', \psi')$ , i. e., for some  $\phi' \in \Phi'$  we have  $\phi' \equiv \neg \psi$ .

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# **Proposition**

Let  $\Delta = (K, A)$  be a knowledge base. Then  $\phi \triangleright_{\Delta} \psi$  if and only if every stable extension E of  $F(K \cup \{\phi\}, A)$  contains an argument with conclusion  $\psi$ .

Claudette Cayrol. "On the Relation between Argumentation and Non-monotonic Coherence-Based Entailment". In: Proceedings of the Fourteenth International Joint Conference on Artificial Intelligence, IJCAI 95, Montréal Québec, Canada, August 20-25 1995, 2 Volumes. Morgan Kaufmann, 1995, pp. 1443–1448

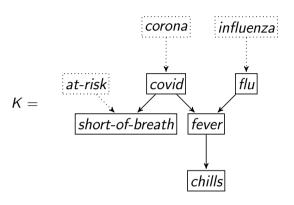
## Example

# **Example**

Consider the causal knowledge base  $\Delta = (K, A)$ , where

```
K = \begin{array}{cccc} covid & \leftrightarrow & corona \\ flu & \leftrightarrow & influenza \\ fever & \leftrightarrow & covid \lor flu \\ chills & \leftrightarrow & fever \\ short-of-breath & \leftrightarrow & covid \land at-risk \\ \end{array}
```

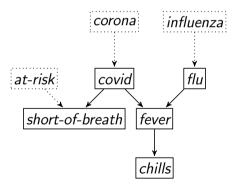
$$A = \{at\text{-}risk, \neg at\text{-}risk, \neg corona, \neg influenza\}$$



$$A = \{\textit{at-risk}, \neg \textit{at-risk}, \neg \textit{corona}, \neg \textit{influenza}\}$$

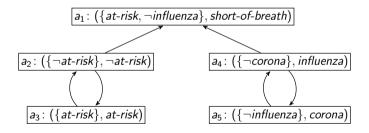
# **Example**

fever  $\sim_{\Delta}$  short-of-breath



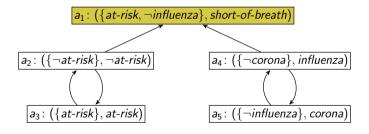
## **Example**

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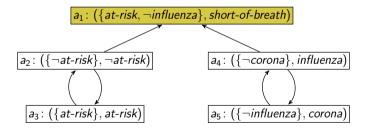
## **Example**

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fever  $\sim_{\Delta}$  short-of-breath



⇒ Given fever, shortness of breath is possible but not necessary.

# Counterfactual Reasoning

given  $\phi, \text{ if } \textit{v} \text{ had been } \textit{x} \text{ then } \psi \text{ would be true}$ 

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#### **Definition**

The *twin model* for a causal model K is the causal model  $K^*$  defined by

$$K^* = K \cup \{(v^* \leftrightarrow \phi^*) \mid (v \leftrightarrow \phi) \in K\}$$

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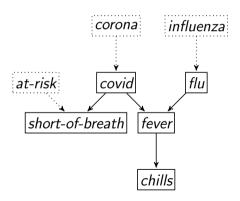
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## **Example**

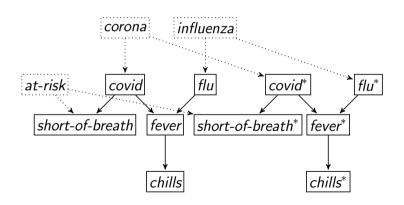
Would the patient have had *fever* if we had administered a covid vaccine (i.e., if *covid* had been false)?

$$\mathit{fever} \hspace{0.2em} \hspace$$

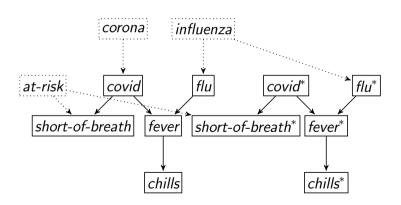
$$\mathit{fever} \hspace{0.2em} \hspace$$











$$\mathit{fever} \hspace{0.2em} \hspace$$

$$\boxed{a_1 \colon (\{\neg corona\}, fever^*)} \longleftarrow \boxed{a_2 \colon (\{\neg influenza\}, corona)}$$
$$\boxed{a_3 \colon (\{\neg corona\}, influenza)}$$

$$\mathit{fever} \hspace{0.2em} \hspace$$

$$a_1: (\{\neg corona\}, fever^*)$$

$$a_2: (\{\neg influenza\}, corona)$$

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## **Example**

$$\mathit{fever} \hspace{0.2em} \hspace$$

$$a_1: (\{\neg corona\}, fever^*)$$

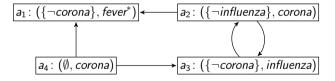
$$a_2: (\{\neg influenza\}, corona)$$

$$a_3: (\{\neg corona\}, influenza)$$

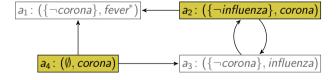
⇒ The patient may or may not have had fever, had we administered a covid vaccine.

$$\textit{fever} \land \textit{short-of-breath} ~ {\sim_{\Delta^*_{[\textit{covid}^* = \bot]}}} ~ \textit{fever}^*$$

$$fever \land short\text{-}of\text{-}breath \hspace{0.2em} \hspace{0.2em} \hspace{0.2em} \swarrow_{\Delta^*_{[covid^*=\bot]}} fever^*$$

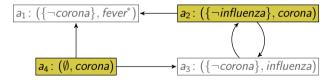


$$\textit{fever} \land \textit{short-of-breath} \, \middle|_{\Delta^*_{[\textit{covid}^* = \bot]}} \, \textit{fever}^*$$



### **Example**

$$\textit{fever} \land \textit{short-of-breath} ~ {\sim_{\Delta^*_{[\textit{covid}^* = \bot]}}} ~ \textit{fever}^*$$



⇒ The patient would not have had fever, had we administered a covid vaccine.

### Conclusion and Future Work

- ▶ We defined a model to transform causal knowledge bases into Dung-style AFs.
- ► The twin model method provides an alternative mechanism for answering counterfactual queries.
- ▶ The constructed AF can be used to provide argumentative explanations.

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## **Open Questions**

▶ Representing uncertain causal relations via probabilistic argumentation methods.