

# Causal Reasoning by Alexander Bochman

Florian Endel

Artificial Intelligence Group,  
University of Hagen, Germany

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**1** Introduction & Background

2 Logical Formal Theory of Causal Reasoning

3 Closing remarks

Relevant for my presentation:

- ▶ [1] *A Logical Theory of Causality*. 2021
- ▶ [2] **“Causal reasoning from almost first principles”**. 2024

Early work:

- ▶ [3] “A logic for causal reasoning”. 2003
- ▶ [4] “A causal approach to nonmonotonic reasoning”. 2004

Exploring various aspects:

- ▶ [5] “Actual Causality in a Logical Setting”. 2018
- ▶ [6] “Default Logic as a Species of Causal Reasoning”. 2023
- ▶ [7] “An Inferential Theory of Causal Reasoning”. 2023

- ▶ "The interpretation of probability as propensity leads people to base their judgments of likelihood primarily on causal considerations, and to ignore information that does not have causal significance." [8]
- ▶ "The possibility to learn causal relationships from raw data has been on philosophers' dream list since the time of Hume (1711-1776)" [9]
- ▶ "causal intent, inference, implications, and recommendations — is common" (in the observational health literature) [10]

... but how to formalize causality in formal logic ?

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[8] Kahneman. *Thinking, Fast and Slow*. 2013

[9] Pearl. *Causality: Models, Reasoning, and Inference*. 2000

[10] Haber et al. "Causal and Associational Language in Observational Health Research: A Systematic Evaluation". 2022

**1** Introduction & Background

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- ▶ Causation is multifaceted: no formalization of concepts
- ▶ Approach: "investigate important variants of causal reasoning"[2]
- ▶ Definition of systems of reasoning:
  - ▶ Language consists of set of (causal) inference rules defined on a set of propositions
  - ▶ Semantics : valuations of propositions (with causal rules)
  - ▶ Causal Inference : formal derivations preserving rational semantics
- ▶ Distinction: rational semantics vs. causal theory
- ▶ Inherently nonmonotonic

- ▶ ... general logical formalism of causal reasoning [5]
- ▶ 2 layers: nonmonotonic semantics + logics of causal rules [4]

## Causal Language

- ▶ Built on top of classical logic
- ▶ Language: set of causal rules
- ▶ Underlying language  $L$ : set of proposition

## Basics: classical propositional language [1, p. 79]

- ▶  $\wedge, \vee, \neg, \rightarrow, t, f$ : classical connectives & constants
- ▶  $\text{Th} (\vdash)$ : Syntactic *provability*, based on formal proof rules.
- ▶  $\models$ : Semantic *entailment*, based on truth in all models.
- ▶  $p, g, r$  finite sets of propositions;  $A, B, C$  classical propositions

## Causal rule

- ▶ Causal binary relation / causal rule:  $a \Rightarrow B$  : *a causes B*
- ▶ Causal theory  $\Delta$ : arbitrary set of causal rules

## Principles of Acceptance

- ▶ Causal Acceptance Principle: *B is accepted iff  $a \Rightarrow B$ , where all  $A$  in  $a$  are accepted*
- ▶ **Preservation** Principle: If all propositions in  $a$  are accepted, and  $a$  causes  $B$ , then  $B$  should be accepted.
- ▶ Principle of **Sufficient Reason**: Any proposition should have a cause for its acceptance.



## Valuation on propositions for describing semantics

- ▶ function  $v \in \{0, 1\}^L$
- ▶  $v(A) = 1$ : proposition  $A$  is accepted ('taken-true')
- ▶  $v(A) = 0$ : non-acceptance is not rejection of  $A$
- ▶ rejection:  $v(\neg A) = 1$
- ▶  $\Delta(u) = \{B \mid a \Rightarrow B \in \Delta, a \subseteq u\}$
- ▶ Causal model: fixed point of (accepted) propositions  
 $v = \Delta(v)$
- ▶ Least model  $\Delta()$ : smallest model
  - ▶  $u_0 = \emptyset, \quad u_1 = \Delta(u_0), \quad u_2 = \Delta(u_1), \dots$
- ▶ Rational Semantics: set of all causal models of a theory

## Causal theory

Rained  $\Rightarrow$  Grasswet

Sprinkler  $\Rightarrow$  Grasswet

Rained  $\Rightarrow$  Streetwet

Formal derivations "*metainferences*" among causal rules that always preserve the rational semantics:

- ▶ **Monotonicity:** If  $a \Rightarrow A$  and  $a \subseteq b$ , then  $b \Rightarrow A$ .
- ▶ **Cut:** If  $a \Rightarrow A$  and  $a, A \Rightarrow B$ , then  $a \Rightarrow B$ .

## Reflexivity and Causal assumptions

- ▶  $A \Rightarrow A$  does **not** hold by default
  - ▶  $A \Rightarrow A$  is a self-evident proposition that does not require further justification for its acceptance
- 
- ▶  $\Rightarrow_{\Delta}$ : all causal rules that are derivable from  $\Delta$ ;  $a \Rightarrow_{\Delta} B$
  - ▶ Any causal theory  $\Delta$  is semantically equivalent to  $\Rightarrow_{\Delta}$

- ▶  $\mathcal{C}(u)$  is the set of propositions ( $A$ ) that are caused by a set  $u$ :  
 $\mathcal{C}(u) = \{A \mid u \Rightarrow A\}$
- ▶ Similar to the derivability operator  $Th(u)$
- ▶ **Monotonicity**: adding more propositions to the set  $u$ , the set of consequences  $\mathcal{C}(u)$  only grows or stays the same – it cannot shrink.
  - ▶ If  $u \subseteq v$ , then  $\mathcal{C}(u) \subseteq \mathcal{C}(v)$
- ▶ **Deductive Closure**:  $\mathcal{C}(u)$  always results in a *deductively closed set*: all logical consequences of the elements in  $\mathcal{C}(u)$  are also included in  $\mathcal{C}(u)$ 
  - ▶  $\mathcal{C}(u) = Th(\mathcal{C}(u))$
  - ▶ e.g.: if the rain ( $u$ ) causes the street to be wet ( $\mathcal{C}(u)$ ), it also causes the logical consequences  $Th(\mathcal{C}(u))$  (e.g., a slippery street)
- ▶ **Transitivity**: if  $A \Rightarrow B$  and  $B \Rightarrow C$ , then  $A \Rightarrow C$

### Non-inclusivity of $\mathcal{C}$

$u$  is not necessarily a subset of  $\mathcal{C}(u)$ :  $u \not\subseteq \mathcal{C}(u)$

- ▶ set of propositions that are caused by  $u$  (i.e.,  $\mathcal{C}(u)$ ) do not always contain  $u$  itself
- ▶ just because a proposition is in  $u$  it must not appear in the set of consequences caused by  $u$
- ▶ Non-inclusivity related to (non-) reflexivity
- ▶ Contrast to derivability operator Th: original set is always included in the set of derivable propositions

### Example: $\mathcal{C}(\text{rain})$

a specific raining event ( $u$ ) has a causal consequences ( $\mathcal{C}(u)$ , e.g. a wet street), but does not necessarily cause itself (but perhaps other raining events  $u'$ )

## Non-idempotence of $\mathcal{C}$

- ▶  $\mathcal{C}(\mathcal{C}(u)) \neq \mathcal{C}(u)$
- ▶ a set of propositions that are caused by  $u$  ( $\mathcal{C}(u)$ ) might cause another set of propositions  $\mathcal{C}(\mathcal{C}(u))$
- ▶ causation propagates across several steps (cascading effects)

## Example: $u = \{\text{rain}\}$

- ▶ 1. Step: Direct causal effects of  $u$ :  $\mathcal{C}(u) = \{\text{wet street}\}$ 
  - ▶ causes logical consequences  $Th(\mathcal{C}(u))$  directly: slippery street
- ▶ 2. Step: Effects of  $\mathcal{C}(u) = \{\text{slippery street}\}$ :
  - ▶  $\mathcal{C}(\mathcal{C}(u)) = \{\text{people slip on the street, traffic slows down}\}$
- ▶ 3. Step: Effects of  $\mathcal{C}(\mathcal{C}(u))$ :
  - ▶  $\mathcal{C}(\mathcal{C}(\mathcal{C}(u))) = \{\text{people get hurt, cloth get wet}\}$

- ▶ A **causal theory** is a set of causal rules (conditionals) that define the causal behavior within a certain framework.
- ▶  $\mathcal{C}_\Delta$ : For any causal theory  $\Delta$ , there is a least production relation, that includes all causal rules derivable from  $\Delta$ .

If theory  $\Delta$  states:

$$\{A \Rightarrow B, B \Rightarrow C\},$$

then for a premise set  $u = \{A\}$ , one can systematically derive

$$\mathcal{C}_\Delta(u) = \{B, C\},$$

representing both intermediate and direct effects triggered by  $A$ .

# Causal Inference ( $\Rightarrow$ ) vs. Deductive Consequence ( $\vdash_{\Rightarrow}$ )

- ▶  $\Rightarrow$ : reasoning based on *cause-and-effect* relationships
- ▶  $\vdash_{\Rightarrow}$ : propositional theory about the consequence relation (rules + results).
- ▶  $\Rightarrow$  has the same propositional theories as  $\vdash_{\Rightarrow}$
- ▶ **But**: causal reasoning retains extra causal information beyond what is encoded in the propositional theories.

## Example

- ▶ *Rained*  $\Rightarrow$  *Grasswet*, *Sprinkler*  $\Rightarrow$  *Grasswet* lead to *Grasswet*
- ▶ **But** the specific causal mechanisms (rain vs. sprinkler) are part of the causal model, not captured in the resulting propositional theory.



- ▶ Logically equivalent  $\Delta$  are semantically equivalent
- ▶ Reverse does not hold!

- ▶ Two theories:

$$\Delta = \{A \Rightarrow B\}, \Phi = \{A \Rightarrow C\}$$

- ▶ different, determine same rational semantics  $\Delta() = \Phi() = \emptyset$
- ▶ add causal rule  $A \Rightarrow A$  to both
- ▶  $\Delta() = \{A, B\}, \quad \Phi() = \{A, C\}$

- ▶ **Axiom:**  $\emptyset \Rightarrow A$
- ▶ **Causal Assumption:**  $A \Rightarrow A$
- ▶ (Relationship to abductive reasoning [1])

## Original Example: Rained, Sprinkler, Streetwet

- ▶ by itself, this causal theory has a single empty causal model  $\emptyset$
- ▶ add assumptions:  $Rained \Rightarrow Rained$     $Sprinkler \Rightarrow Sprinkler$
- ▶ Rational semantics of this causal theory: three causal models
- ▶  $\{Rained, Grasswet, Streetwet\}$ ,  $\{Sprinkler, Grasswet\}$ ,  
 $\{Rained, Sprinkler, Grasswet, Streetwet\}$

- ▶ Aim: integrating causal reasoning into a more comprehensive reasoning system
- ▶ **Classical entailment** (formal logical reasoning): integral part of the causal system
- ▶ Supraclassical: extends and includes classical entailment, to naturally incorporate classical logic
- ▶ "Causal reasoning is not a replacement or competitor of logical (deductive) reasoning, but its complement (or extension)" [1]
- ▶ (*Strengthening*) If  $A \models B$  and  $B \Rightarrow C$ , then  $A \Rightarrow C$ ;
- ▶ (*Weakening*) If  $A \Rightarrow B$  and  $B \models C$ , then  $A \Rightarrow C$ ;
- ▶ (*And*) If  $A \Rightarrow B$  and  $A \Rightarrow C$ , then  $A \Rightarrow (B \wedge C)$ ;
- ▶ (*Truth and Falsity*)  $t \Rightarrow t$ ;  $f \Rightarrow f$ .

- ▶ **Default causal theory:** pair  $(\Delta, \mathcal{D})$ ,  $\mathcal{D}$  a subset of causal assumptions
- ▶ **Nonmonotonicity**
  - ▶ Context-Sensitive Causal Reasoning
  - ▶ Causal acceptance is directional
- ▶ **Structural Equation Models:** representation in the causal calculus
- ▶ Counterfactual Equivalence: related to SEM
- ▶ Negative Causal Completion: negation as default ( $\neg p \Rightarrow \neg p$ )

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- ▶ Explainable AI [2]
- ▶ Causal attribution (actual causality) in legal theory [5]

Bochman's causal reasoning...

- ▶ Unified approach: integrate Pearl's approach to causation, classical entailment within causal reasoning
- ▶ Nonmonotonic reasoning
- ▶ Asymmetry in Semantics
- ▶ Roots in Historical Reasoning
- ▶ Connection to Inferentialism (rule-based vs. representational)

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## 4 Appendix

- Bibliography

## 5 Spare Slides

- Classical propositional language

## 6 Example slides

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## 6 Example slides

- [1] Alexander Bochman. *A Logical Theory of Causality*. The MIT Press, Aug. 17, 2021. ISBN: 978-0-262-36620-5. DOI: [10.7551/mitpress/12387.001.0001](https://doi.org/10.7551/mitpress/12387.001.0001).
- [2] Alexander Bochman. “Causal reasoning from almost first principles”. In: *Synthese* 203.1 (Jan. 4, 2024), p. 19. ISSN: 1573-0964. DOI: [10.1007/s11229-023-04442-6](https://doi.org/10.1007/s11229-023-04442-6).
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- [11] Judea Pearl. “Embracing causality in default reasoning”. In: *Artificial Intelligence* 35.2 (June 1, 1988), pp. 259–271. ISSN: 0004-3702. DOI: 10.1016/0004-3702(88)90015-X.
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- [13] Hector Geffner. “Causal theories for nonmonotonic reasoning”. In: *Proceedings of the eighth National conference on Artificial intelligence - Volume 1. AAAI'90*. Boston, Massachusetts: AAAI Press, July 29, 1990, pp. 524–530. ISBN: 978-0-262-51057-8.
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- [15] D. Dash. “Causal Logic Models”. In: 2012.

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Some of the most influential authors:

- ▶ [11] Pearl. “Embracing causality in default reasoning”. 1988
- ▶ [9] Pearl. *Causality*. 2000
- ▶ [12] Lifschitz. “On the logic of causal explanation”. 1997
- ▶ [13] Geffner. “Causal theories for nonmonotonic reasoning”. 1990
- ▶ [14] Turner. “A logic of universal causation”. 1999
- ▶ [15] Dash. “Causal Logic Models”. 2012

... and many others

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**Definition:**  $Th(\vdash)$  represents syntactic provability, meaning a proposition  $A$  can be formally derived using axioms and formal inference rules. Provability is about whether a proposition  $A$  can be formally derived from a set of axioms or premises  $\Gamma$  using a given set of syntactic proof rules within a formal system.

**Key Idea:**

- ▶ A proposition  $A$  is provable (written as  $Th \vdash A$  if there is a formal proof sequence for  $A$ ).
- ▶ Relies entirely on the structure and rules of a formal system (e.g., axioms, inference rules like Modus Ponens).
- ▶ Does not involve interpretations or models; it's purely rule-based reasoning.
- ▶ Following strict rules to deduce a conclusion.

**Definition:**  $\models$  represents semantic entailment, meaning a proposition  $A$  is true in any model where the assumptions  $\Gamma$  hold. Entailment is about whether  $A$  is true in every possible model (interpretation) where a set of premises  $\Gamma$  is true.

**Key Idea:**

- ▶ Determines logical consequence based on truth in all models, not formal derivation.
- ▶ Like checking universal truths: something is true in all possible worlds where the premises hold.
- ▶ Depends on truth values assigned to propositions in all models of the logic. Assesses whether a logical proposition holds in a model-theoretic sense.

**Definition:** Propositional atoms are the smallest, indivisible units of logic, representing atomic facts.

**Example:**

$p = \text{"It is raining"}$ ,  $g = \text{"The grass is wet"}$ ,  $r = \text{"It is windy"}$ .

No further breakdown is possible for these logical units.

**Definition:** Classical propositions are statements formed by combining propositional atoms with logical connectives ( $\wedge, \vee, \neg, \rightarrow$ ) or truth constants ( $t, f$ ).

**Example:**

$$A = p \wedge g, \quad B = p \vee \neg r, \quad C = \neg g \rightarrow r.$$

These statements are more complex and their truth values depend on logical interpretation.

- ▶ Syntactic Provability ( $\vdash$ ): Formal derivation based on rules.
- ▶ Semantic Entailment ( $\models$ ): Truth across all models.
- ▶ Propositional Atoms: Indivisible units ( $p, g, r$ ).
- ▶ Classical Propositions: Composite logical statements ( $A, B, C$ ).

These concepts bridge syntactic and semantic aspects of logic, fundamental to understanding relationships in formal reasoning systems.

- ▶ **Provability:**  $Th \vdash Grasswet$   
Derive *Grasswet* syntactically using axioms like  $Rained \rightarrow Grasswet$ .
- ▶ **Entailment:**  $Rained \models Grasswet$   
*Grasswet* is true in every model where *Rained* is true.
- ▶ **Propositional Atoms:** *Rained*, *Grasswet*, and *Streetwet* are indivisible facts used to construct more complex statements.
- ▶ **Classical Proposition:**  $Rained \rightarrow (Streetwet \vee Grasswet)$   
A composite logical statement describing causal or correlative relationships.

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## 6 Example slides

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## Block

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- ▶ This is an item in a block.

## Block

This is an alert block.

- ▶ This is an item in an alert block.

## Block

This is an example block.

- ▶ This is an item in an example block.