

Causal Graph Justifications of Stable Models

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Corunna, Spain

February 24th, 2015

Causality and Knowledge Representation

- For **Knowledge Representation**, not just deriving conclusions but sometimes we require **explanations**
- **Causality**: is a quite common concept in **human daily discourse**. **Present in** (chronologically or physically) **distant cultures**.
- What “**A** has caused **B**” actually means?

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- What “**A** has caused **B**” actually means?
 - ▶ Sufficient cause
 - ▶ Necessary cause
 - ▶ Actual or contributory cause

Joint interaction

Example

- There is a law asserts that *driving drunk* is *punishable*.
- Suppose that some person drove drunk.

Take the **logic program** consisting of one rule and two **labelled facts**

punish \leftarrow *drive, drunk* *d* : *drive* *k* : *drunk*

- **Joint interaction** of multiple events.
The cause formed by “ $\{d, k\}$ ” together has caused *punish*”.

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The cause formed by “ $\{d, k\}$ ” together has caused *punish*”.
- Two kinds of causal rules:
 - ▶ **Unlabelled rules**: tracing them is irrelevant for causal purposes.
 - ▶ **Labelled rules**: keep track of possible ways to derive an effect.

Labels

- We may want to **keep track** of involved rules and not only facts:

Example

- Law ℓ asserts that *driving drunk* is *punishable* with imprisonment.
- The execution e of a sentence establishes that people who are *punished* are *imprisoned*.
- Suppose that some person drove drunk.

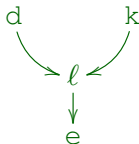
ℓ : *punish* \leftarrow *drive, drunk*

d : *drive*

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e : *prison* \leftarrow *punish*

- We get a **cause** in the form of a label **graph**



Main ideas

- **Multi-valued** semantics for logic programs: each true atom will be associated to a set of justifications (**causal graphs**)
- Accordingly, **falsity = lack of justification**.
 - ▶ This coincides with the informal reading for **default negation**:
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- Causes must be **non-redundant**.
 - ▶ Some causes will be **stronger** than others.
 - ▶ This allows us defining a lattice and algebraic operations
+ (alternative causes), * (joint causation) and · (rule application).

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 - ▶ This allows us defining a lattice and algebraic operations
 $+$ (alternative causes), $*$ (joint causation) and \cdot (rule application).
- Important result: **semantically obtained causal values** correspond to (non-redundant) **syntactic proofs** using the program rules!

Outline

- 1 Motivation and examples
- 2 Causes as graphs**
- 3 Positive programs
- 4 Default negation
- 5 Queries about causality
- 6 Conclusions and future work

Causal Graphs

Definition

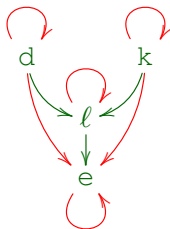
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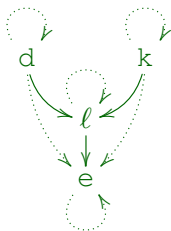


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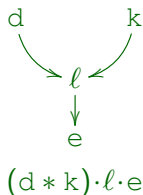
Causal Graphs

- G^* is the transitive and reflexive closure of G
- Product $G * G' \stackrel{\text{def}}{=} (G \cup G')^*$
- Application $G \cdot G' \stackrel{\text{def}}{=} \text{graph with vertices } V \cup V' \text{ and edges } E \cup E' \cup \{ (x, y) \mid x \in V, y \in V' \}$
- Atomic graphs l stands for $\langle \{l\}, \{(l, l)\} \rangle$



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- Atomic graphs l stands for $\langle \{l\}, \{(l, l)\} \rangle$ 
- Any causal graph can be built from **product**, **application** and **atomic graphs**. Example:



Causal Graphs

Definition

A causal graph G is **sufficient** for (or **weaker** than) another causal graph G' , written $G \leq G'$, when $G \supseteq G'$.

- Notice that **direction is switched**: the smaller the graph, the **stronger** the cause!

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- The empty graph $\langle \emptyset, \emptyset \rangle$ is the top element, denoted by 1 .
 - ▶ stands for absolute truth, and assigned to \top .
 - ▶ 1 is the $*$ product and \cdot application **identity** $t * 1 = t$ and $t \cdot 1 = 1 \cdot t = t$

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- We add a bottom element 0 ,
 - ▶ weaker than any causal graph $0 < G$ for all G ,
 - ▶ stands for **false**,
 - ▶ 0 is the $*$ and \cdot application **annihilator** $t * 0 = 0$ and $t \cdot 0 = 0 \cdot t = 0$

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Positive programs

- Syntax: as usual plus an (optional) **rule label**

$$t : H \leftarrow B_1, \dots, B_n$$

with H, B_i atoms and t can be a label $t = \ell$ or $t = 1$.

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Definition (Causal model)

A **causal model** of P is an interpretation such that, for each rule:

$$(\mathcal{I}(B_1) * \dots * \mathcal{I}(B_n)) \cdot t \leq \mathcal{I}(H)$$

Alternative causes (symmetrical overdetermination)

Example

- A second law m specifies that *resisting* to authority is *punishable*.
- Suppose that some person drove drunk and resisted to authority.

l : *punish* \leftarrow *drive, drunk*

d : *drive*

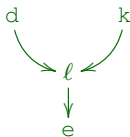
k : *drunk*

e : *prison* \leftarrow *punish*

m : *punish* \leftarrow *resist*

r : *resist*

- Two equally valid **alternative causes**



$(d * k) \cdot l \cdot e$



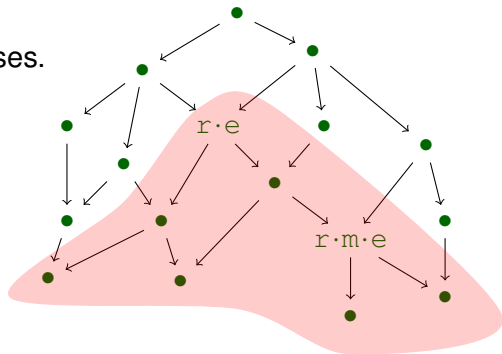
$r \cdot m \cdot e$

Alternative causes: Addition

- addition (+) represents alternative causes

$$\mathcal{I}(\textit{punish}) = (d * k) \cdot \ell + r \cdot m \cdot e$$

- Causal values are ideals of causal graphs. (+) corresponds to the union (\cup) of ideals.
- Disregard redundant causes.



Alternative causes

Theorem

$\langle \mathbf{V}_{Lb}, +, *, \cdot \rangle$ is the free algebra generated by labels Lb . Operations $*$ and $+$ are the meet and join of a completely distributive lattice.

Associativity

$$\begin{array}{l} t + (u+w) = (t+u) + w \\ t * (u*w) = (t*u) * w \end{array}$$

Commutativity

$$\begin{array}{l} t + u = u + t \\ t * u = u * t \end{array}$$

Absorption

$$\begin{array}{l} t = t + (t*u) \\ t = t * (t+u) \end{array}$$

Distributive

$$\begin{array}{l} t + (u*w) = (t+u) * (t+w) \\ t * (u+w) = (t*u) + (t*w) \end{array}$$

Identity

$$\begin{array}{l} t = t + 0 \\ t = t * 1 \end{array}$$

Annihilator

$$\begin{array}{l} 1 = 1 + t \\ 0 = 0 * t \end{array}$$

Alternative causes

- More specific are the (\cdot) application equations

$$\frac{\textit{Associativity}}{t \cdot (u \cdot w) = (t \cdot u) \cdot w}$$

$$\frac{\textit{Addition distributivity}}{t \cdot (u + w) = (t \cdot u) + (t \cdot w)} \\ (t + u) \cdot w = (t \cdot w) + (u \cdot w)$$

$$\frac{\textit{Identity}}{t = t \cdot 1} \\ t = 1 \cdot t$$

$$\frac{\textit{Annihilator}}{0 = t \cdot 0} \\ 0 = 0 \cdot t$$

$$\frac{\textit{Absorption}}{t = t + u \cdot t \cdot w} \\ u \cdot t \cdot w = t * u \cdot t \cdot w$$

- l is a label, c , d and e terms without $(+)$

$$\frac{\textit{Label idempotence}}{l \cdot l = l}$$

$$\frac{\textit{Product distributivity}}{c \cdot (d * e) = (c \cdot d) * (c \cdot e)} \\ (c * d) \cdot e = (c \cdot e) * (d \cdot e)$$

$$\frac{\textit{Transitivity}}{c \cdot d \cdot e = (c \cdot d) * (d \cdot e) \text{ with } d \neq 1}$$

Positive programs

Definition (Direct consequences)

$$T_P(\mathcal{I})(p) \stackrel{\text{def}}{=} \sum \{ (\mathcal{I}(B_1) * \dots * \mathcal{I}(B_n)) \cdot t \mid (t : p \leftarrow B_1, \dots, B_n) \in P \}$$

Theorem (Analogous to standard LP)

Let P be a (possibly infinite) positive logic program with n causal rules.

- (i) $\text{lfp}(T_P)$ is the least model of P ,
- (ii) $\text{lfp}(T_P) = T_P \uparrow^\omega (\mathbf{0})$, and
- (iii) iteration ends in *finite* steps when P is finite $\text{lfp}(T_P) = T_P \uparrow^n (\mathbf{0})$.

Theorem

Removing all labels we get the traditional (two-valued) least model.

Positive programs

- Positive programs have a least model.

$$\mathcal{I}(\textit{prison}) = (d * k) \cdot \ell \cdot e + r \cdot m \cdot e$$

- If we remove all labels, then it corresponds to the standard least model.

$$\mathcal{I}(\textit{prison}) = 1$$

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- Each **subterm with no sums** is a **cause**. But what do causal values really capture?
 - ▶ syntactic proofs?
 - ▶ some proofs? all proofs?
- Notice we have **not used syntactic information!**

Positive programs

Theorem

The causal value of an atom in the least model *exactly corresponds to all its possible (non-redundant) proofs.*

$(d * k) \cdot \ell \cdot e$

$$\frac{\frac{\overset{T}{drive} (d)}{\quad} \quad \frac{\overset{T}{drunk} (k)}{\quad}}{punish (l)} (e)$$

$r \cdot m \cdot e$

$$\frac{\overset{T}{resist} (r)}{punish (m)} (e)$$

Outline

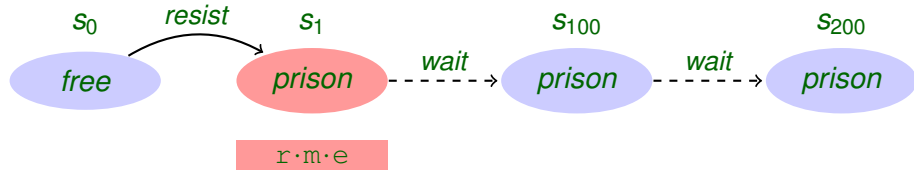
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 - ▶ Suppose now that we add **time** to our running example and we are imprisoned by *resist* at situation s_1 , then

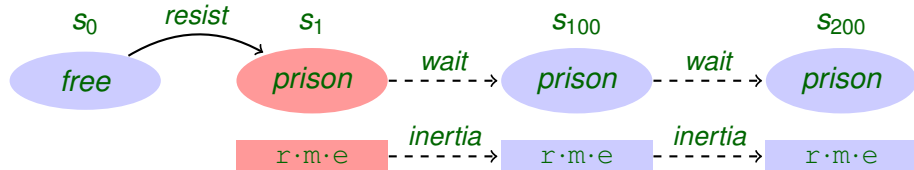


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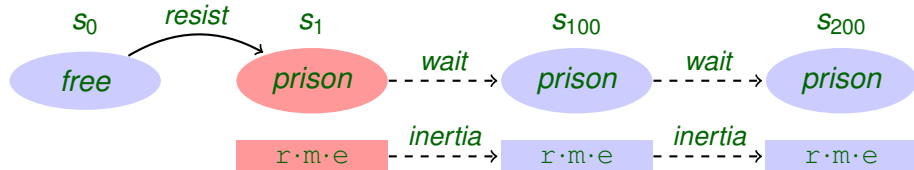


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- Causal values persist by inertia. We **disregard** explanations for *not* being *free* along that period!

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m : *punish* \leftarrow *resist*

e : *prison* \leftarrow *punish, not abnormal*

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- If we assume $\mathcal{I}(\textit{abnormal}) = 0$ (false).

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Default negation

- we can flexibly add **exceptions**

abnormal \leftarrow *pardon*
abnormal \leftarrow *revoke*
abnormal \leftarrow *diplomat*

- If we assume to be a *diplomat*, then $\mathcal{I}(\textit{abnormal}) = 1$ (true).

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Theorem

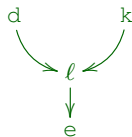
For each (standard) two-valued stable model there is (exactly one) corresponding causal stable model and vice versa.

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Sufficient Cause

- Why are we in prison?
 - ▶ *sufficient*($X, prison$)?, X should be a minimal explanation



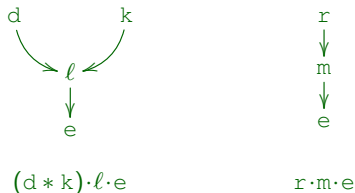
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$r \cdot m \cdot e$

Sufficient Cause

- Why are we in prison?
 - ▶ $\text{sufficient}(X, \text{prison})?$, X should be a minimal explanation



- ▶ Was $d * k * \text{chew}$ sufficient to cause it?
- ▶ $\text{sufficient}(d * k * \text{chew}, \text{prison})$ should hold, despite of lack of minimality

Sufficient Cause

- Given a causal graph G
 - ▶ G is a sufficient explanation for p iff $G \leq I(p)$
 - ▶ G is a sufficient cause for p iff G is a subgraph-minimal sufficient explanation for p

Sufficient Cause

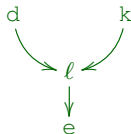
- Given a causal graph G
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- Complexity (complete results)

	positive	well founded	answer set	
			(brave)	(cautions)
entailment	P	P	NP	coNP
explanation	P	P	NP	coNP
cause	P	P	NP	coNP

- same complexity than entailment in standard LP

Necessary Cause

- Why are we in prison?
 - ▶ What has been necessary to cause it?



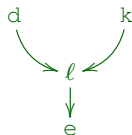
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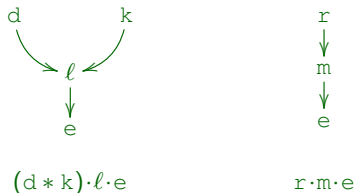


$$r \cdot m \cdot e$$

- ▶ Only the rule e has been necessary.

Necessary Cause

- Why are we in prison?
 - ▶ What has been necessary to cause it?



- ▶ Only the rule e has been necessary.
- ▶ Suppose we do not resit. Then *drive* and *drunk* would have been necessary causes.
- ▶ Suppose we were not drunk. Then *resit* would have been a necessary cause.

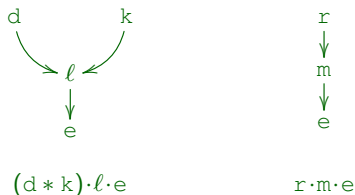
Necessary Cause

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 - G is a necessary cause for p iff G subgraph of all sufficient causes for p and $I(p) \neq 0$
 - G is a necessary cause for p iff $G \geq I(p)$ and $I(p) \neq 0$
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entailment	P	P	NP	coNP
necessary	coNP	coNP	Σ_2^P	coNP

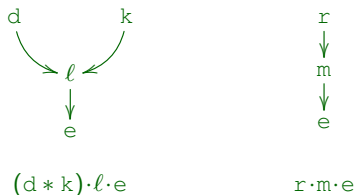
Actual and Contributory Cause

- Why are we in prison?
 - ▶ Actual Cause \approx contingency necessary cause.
 - ▶ There exists a possible world where G is a necessary cause [Pearl 2000, Halpern & Pearl 2001 and 2005].



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- ▶ Contributory cause: Necessary condition in a sufficient cause [Mackie 1965, Wright 1988]

Actual and Contributory Cause

- Given a causal graph G
 - G is a **actual cause** for p iff there exists a sufficient cause G' for p such that $G \subseteq G'$
- Complexity

	positive	well founded	answer set	
			(brave)	(cautions)
entailment	P	P	NP	coNP
actual	\leq NP	\leq NP	\leq NP	$\leq \Pi_2^P$
HP 2001	NP / Σ_2^P			
HP 2005	D_2^P			

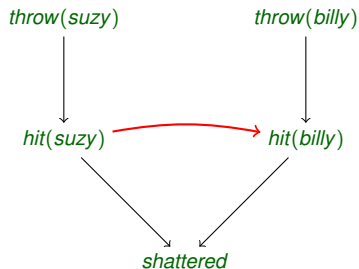
- [Eiter & Lukasiewicz 2001, Aleksandrowicz et. al. 2014]
- $\Sigma_2^P \leq D_2^P \leq \Delta_3^P \leq \Sigma_3^P$
- $\Pi_2^P \leq D_2^P \leq \Delta_3^P \leq \Pi_3^P$

Causality and Knowledge Representation

Example (Lewis2000)

Suzy throws a rock at a bottle. The rock hits the bottle, shattering it. Suzy's friend Billy throws a rock at the bottle a couple of seconds later. Who has caused the bottle to shattered?

$hit(suzy) = throw(suzy)$
 $hit(billy) = throw(billy) \wedge \neg hit(suzy)$
 $shattered = hit(suzy) \vee hit(billy)$



- ▶ Actual Cause in structural equations [Halpern&Pearl2005, Hall2007, Halpern2008, Halpern2014]

Causality and Knowledge Representation

- Suppose that John has also thrown after Billy.

hit(suzy) = *throw(suzy)*

hit(billy) = *throw(billy)* \wedge \neg *hit(suzy)*

hit(john) = *throw(john)* \wedge \neg *hit(suzy)* \wedge \neg *hit(billy)*

shattered = *hit(suzy)* \vee *hit(billy)* \vee *hit(john)*

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$shattered = hit(suzy) \vee hit(billy) \vee hit(john)$

- Change: John has thrown **before** Suzy.

Causality and Knowledge Representation

- Suppose that John has also thrown after Billy.

$$\begin{aligned}hit(suzy) &= throw(suzy) \\hit(billy) &= throw(billy) \wedge \neg hit(suzy) \\hit(john) &= throw(john) \wedge \neg hit(suzy) \wedge \neg hit(billy) \\shattered &= hit(suzy) \vee hit(billy) \vee hit(john)\end{aligned}$$

- Change: John has thrown **before** Suzy.

$$\begin{aligned}hit(suzy) &= throw(suzy) \wedge \neg hit(john) \\hit(billy) &= throw(billy) \wedge \neg hit(suzy) \wedge \neg hit(john) \\hit(john) &= throw(john) \\shattered &= hit(suzy) \vee hit(billy) \vee hit(john)\end{aligned}$$

Causality and Knowledge Representation

- Suppose that John has also thrown after Billy.

$$\begin{aligned}hit(suzy) &= throw(suzy) \\hit(billy) &= throw(billy) \wedge \neg hit(suzy) \\hit(john) &= throw(john) \wedge \neg hit(suzy) \wedge \neg hit(billy) \\shattered &= hit(suzy) \vee hit(billy) \vee hit(john)\end{aligned}$$

- Change: John has thrown **before** Suzy.

$$\begin{aligned}hit(suzy) &= throw(suzy) \wedge \neg hit(john) \\hit(billy) &= throw(billy) \wedge \neg hit(suzy) \wedge \neg hit(john) \\hit(john) &= throw(john) \\shattered &= hit(suzy) \vee hit(billy) \vee hit(john)\end{aligned}$$

- Small changes implies revise the entire model. Problem of tolerance to the elaboration [McCarthy1998]

Causality and Knowledge Representation

Example (Lewis2000)

Suzy throws a rock at a bottle. The rock hits the bottle, shattering it. Suzy's friend Billy throws a rock at the bottle a couple of seconds later. Who has caused the bottle to shattered?

$shattered(T + 1) \leftarrow throws(X, T), not\ shattered(T)$

$throw(suzy, 2)$

$throw(billy, 4)$

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- Inertia axiom

$shattered(T + 1) \leftarrow shattered(T)$

- We may conclude that the bottle is *shattered* at 3, but not who caused it.

Causality and Knowledge Representation

r_1 : *shattered*($T + 1$) \leftarrow *throws*(X, T), *not shattered*(T)

suzy : *throw*(*suzy*, 2)

billy : *throw*(*billy*, 4)

Causality and Knowledge Representation

r_1 : $shattered(T + 1) \leftarrow throws(X, T), not\ shattered(T)$

$suzy$: $throw(suzy, 2)$

$billy$: $throw(billy, 4)$

- We may conclude that the bottle is *shattered* at 3 because



- Note that rule r_1 for $T = 4$ is not in the reduct of the program

Conclusions

- Multi-valued semantics based on (ideals of) causal graphs
- Values capture non-redundant proofs, but with semantic, algebraic operations
- Default negation = absence of cause.
 - ▶ Reduct definition allows defining causal stable models
 - ▶ Allows expressing dynamic defaults (ex: inertia laws)
- Ongoing work:
 - ▶ Studying actual causation.
 - ▶ Adding this causal operators on rule bodies.

Causal Graph Justifications of Stable Models

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Thanks for your attention!

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February 24th, 2015