

Overview

We use structural equation models (SEMs) to interpret counterfactuals

- SEMs represent dependencies between events
- Formally, counterfactuals denote sets of such dependencies
- Intuitively, these can be thought of as possible explanations
- We classify such explanations into four categories, providing a typology of explanatory strategies

Counterfactuals

- We use counterfactuals to talk about things we know to be false
 - If the movie hadn't been so boring, I wouldn't have fallen asleep.
- And to talk about things we're uncertain about
 - If Sam were angry, Pat would have been angry, too. (But I don't know if she was.)
- Counterfactuals describe some relationship between the events
- There are many ways for two events to be related
- (3) If Alice had gone to the party, Bob would have stayed home.
- Does Bob try to avoid Alice? Maybe he's shy
- Maybe he doesn't like her
- Do other circumstances prevent them from attending parties together?
- Maybe they're a couple on a tight budget
- Maybe Bob is actually Alice in disguise
- Does Alice try to avoid Bob?
- Unlike the other scenarios, this one doesn't seem to jive with (3)
- To understand a counterfactual, we have to capture this range of relationships

Informative Counterfactuals Adam Bjorndahl (Cornell University, Mathematics) & Todd Snider (Cornell University, Linguistics) NASSLLI 2014, Maryland

Modeling Relationships

- To capture relationships between events, we use structured possible worlds (Starr 2014) • Worlds are event variables, their values, and **dependencies** between them • Just like truth values, we can use the (non)existence of dependencies to discriminate among worlds • We model these dependencies using Structural Equation Models as formalized in Pearl 2000
- Nodes as events, arrows as dependencies

Key Contrast

We think of agents as *building explanations* rather than evaluating truth in a fully specified model. As such, we take the SEM not as a *given* but as a *goal*.

Graph as given



Start with a fully specified model



Intervene on the model



Evaluate truth in updated model

A Typology of Explanatory Strategies

Direct Cause

Hear

B > H



- A simple direct dependency
- The 'default' assumption A & B covary in the
- Implicated by a counterfactual, can be canceled or strengthened

Additional Cause



- Positing an additional causal factor
- right C conditions
- Ex: Bob dislikes Alice

Rejecting Explanations

• There are many reasons to reject an explanation (including the implicated direct dependency) • It might contradict prior knowledge • It might violate a law of good explanations • e.g. by positing an effect temporally prior to its cause • It might not satisfy the contextual parameter for specificity

Graph as goal



assumptions



Build possible explanations



models

Common Cause



• Positing a shared cause • No direct relation between A and B• Ex: Coin flip to determine who attends

Intermediate Cause



- Positing a mediating factor
- A & B related, but not directly
- Ex: Bob is allergic to Alice's cat

- (3)
- (4)

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Judea Pearl. Causality: Models, Reasoning and Inference. Cambridge Univ Press, 2000. William B. Starr. Structured possible worlds. Ms. Cornell University, 2014.



Mutual Incompatibility

• Some counterfactuals which are individually felicitous are jointly infelicitous • Consider a world where Alice and Bob are married, and live with their young son Doug

> If Alice had gone to the party, Bob would have stayed home. If Alice had gone to the party, Doug would have been home alone.

• Updating with (3) adds a covariance between Aand $\neg B$ to our knowledge base

• Updating with (4) requires that A and B have the same value

• The models compatible with some explanation of (3) are not compatible with any explanation of (4)

Conclusion

e can use structured possible worlds to odel dependencies, and thus counterfactuals ing so provides a natural way to typologize planatory strategies

so yields insight into the mechanism that plains mutually infelicitous counterfactuals

References

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