

Informative Counterfactuals

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- We use counterfactuals all the time:
 - (1) If Alice had gone to the party, Bob would have stayed home.
 - (2) If the movie had been any good, I wouldn't have fallen asleep.
 - (3) If there hadn't been traffic, we would have been on time.
- We can use them to talk about things we know to be false or things we're uncertain about
- (1) typically implies that Alice didn't go to the party and Bob did
- It also communicates some relation between the two events

- There are different ways for the events in (1) to be related
- (1) 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 does not seem to jive with (1)...

Our Proposal

- Counterfactuals denote sets of relationships between events
- We use the mechanics of *structural equation models* to represent these relationships
- This provides a rich set of tools we use to define a typology of *explanatory strategies*
- Our analysis cleanly distinguishes two different kinds of 'backtracking'
- It also provides a principled account of certain kinds of mutually incompatible counterfactuals

Outline

- 1 Overview
- 2 Some preliminaries
- 3 Our proposal
 - Understanding a counterfactual
 - Three explanatory strategies
 - Integrating a counterfactual with our knowledge
- 4 Conclusion

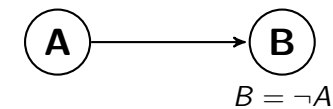
- To capture relationships between events, we use *structured possible worlds* (Starr 2014)
- Worlds are event variables, their values, and **dependencies** between them
 - Just like truth values, we can use the (non)existence of dependencies to discriminate among worlds
- We model these dependencies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

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Structural Equation Models (SEMs)

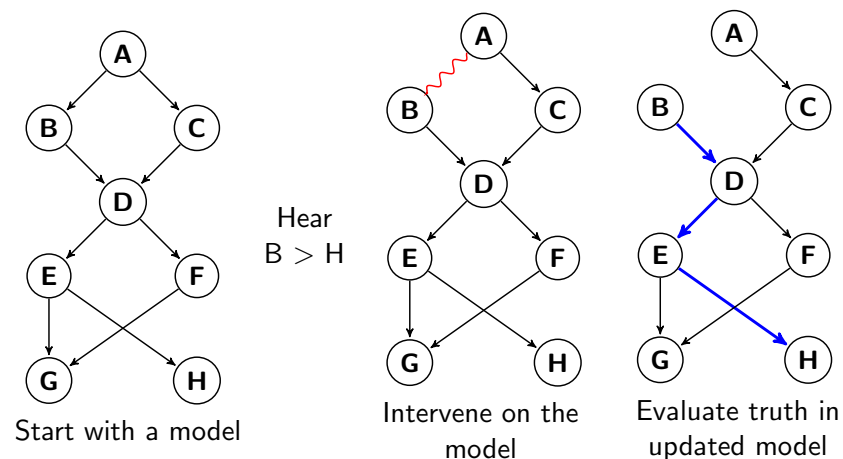
- Allows for the modeling not only of variables but also dependencies
- Models consist of:
 - Nodes Circles Variables/Events
 - Edges Arrows Dependencies
 - Labeled with equations



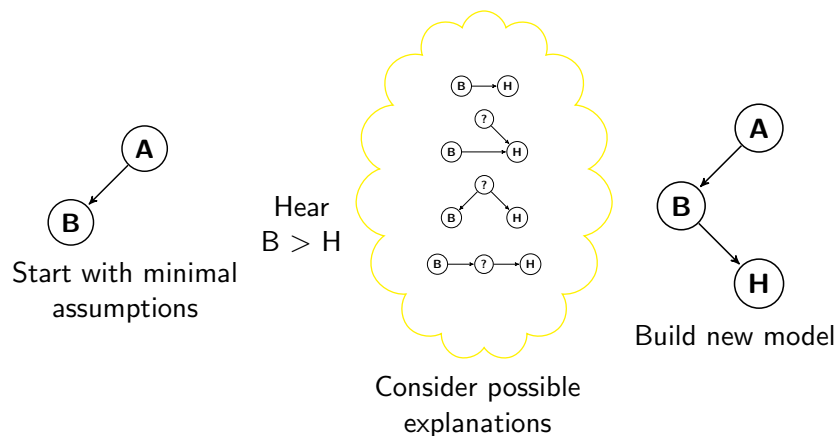
- For convenience and simplicity, our examples are
 - Two-valued
 - Deterministic
- This framework and analysis also handles multi-valued and/or probabilistic systems

- Unlike Pearl, we take the SEM not as a *given* but as a *goal*
- Rather than structures within which to evaluate the truth of a counterfactual, we interpret SEMs as *candidate explanations*
- Counterfactuals denote sets of such explanations

Graph as given



Graph as goal



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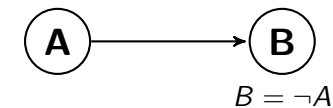
- Counterfactuals assert some degree of covariance between the antecedent and consequent
- (4) If I had pushed this button, the rocket would have launched.
- They implicate a direct (causal) dependence of consequent on antecedent ($C = A$)
- This implicature can be canceled (5) or strengthened (6):
- (5) If I had pushed this button, the rocket would have launched, but pushing this button doesn't directly cause the rocket to launch.
- (6) If I had pushed this button, the rocket would have launched, and (in fact) pushing this button directly causes the rocket to launch.
- Sometimes the direct dependency is problematic

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 that is temporally prior to its cause
 - It might not satisfy the contextual parameter for specificity

- Any of these reasons might make us reject the simple direct dependency of the consequent on the antecedent
 - In other words, reject the $C = A$ edge
- But the counterfactual stipulates some covariance
- Trying to maintain the cooperativity of the speaker's contribution, we search for an explanation to make the counterfactual true
- Three possible ways to deal with this problematic dependence:
 - ADDITIONAL CAUSE
 - COMMON CAUSE
 - INTERMEDIATE CAUSE
- Call these **explanatory strategies**

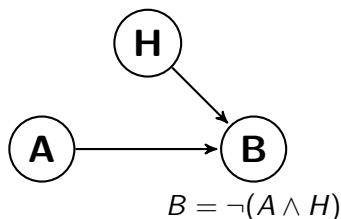
- (1) If Alice had gone to the party, Bob would have stayed home.
- The implicated simple dependency of (1):



- But Alice's attendance doesn't directly cause Bob to be elsewhere
- There are other *explanations*

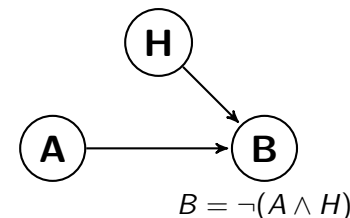
ADDITIONAL CAUSE

- The hearer might suppose that the consequent is dependent not solely on the antecedent but also on some additional cause
- For example, a common interpretation of (1) might lead one to believe that Bob hates Alice
- We can consider Bob's hatred of Alice as an additional node in our model



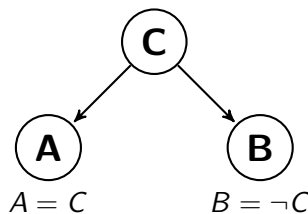
ADDITIONAL CAUSE

- The dependence of B on A is still present, but it's been modified
 - The $B = \neg A$ edge is no longer part of the model
- The antecedent and consequent covary only in the right H-conditions



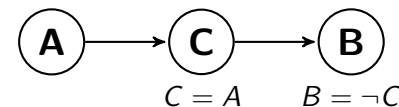
COMMON CAUSE

- The hearer might suppose that the consequent isn't dependent upon the antecedent at all
- Instead, both antecedent and consequent depend on some common cause
- They still covary, but have no interdependence
- For example, imagine that Alice & Bob flip a coin to determine who attends



INTERMEDIATE CAUSE

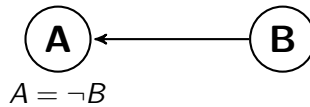
- The hearer might suppose that the consequent depends on the antecedent only by means of some intermediate cause
- The antecedent and consequent still covary, but without positing a direct causal dependency
- For example, imagine that Alice brings her cat wherever she goes, and Bob is deathly allergic to cats



A fourth explanatory strategy?

- Reversing the simple causal relationship also allows the antecedent and consequent to covary

(1) If Alice had gone to the party, Bob would have stayed home.



- This classical *backtracker* has the consequent as the cause
- This model is rejected as an interpretation of (1)
- It's available with a double-auxiliary construction, as in (7)

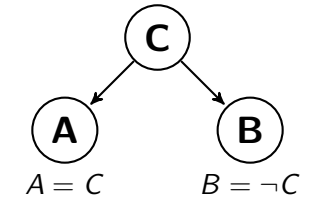
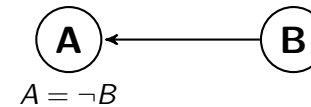
(7) If Alice had gone to the party, Bob would have had to have stayed home.

How do we update with what we've learned?

- Once an acceptable explanation is found, we have to integrate it with our extant body of knowledge
- With structured possible worlds, our knowledge includes not just facts about variables but also dependencies
- We can model our knowledge as one persistent SEM
- When consolidating, we integrate dependencies, not variable values
- Counterfactuals *can* inform us about actual values via presupposition, accommodation
- We don't want to update with Alice's counterfactual attendance

A note on *backtracking*

- Two different things referred to as *backtracking*



- Reversing causal direction
- Classic philosophy literature
- Double-aux environment
- 'Upstream' reasoning
- Recent psychology literature
- Available in (1)

- There are at least two mechanisms involved in consolidation

1 Addition

- Extending the graph
- Possibly add new nodes
- Add new dependencies among nodes

2 Expansion

- Looking deeper into the internal mechanism of a single node
- Explode one node into multiple nodes
- Retains incoming/outgoing dependencies of the original node

- After consolidation, deduce values of new nodes, if necessary

- Consider a world where Alice and Bob are married, and live with their young son Doug
- (1) If Alice had gone to the party, Bob would have stayed home.
- (8) If Alice had gone to the party, Doug would have been home alone.
- (1) and (8) are each felicitous individually
- A felicitous utterance of one precludes a felicitous utterance of the other
- Any account of how we update our knowledge with counterfactuals should explain this

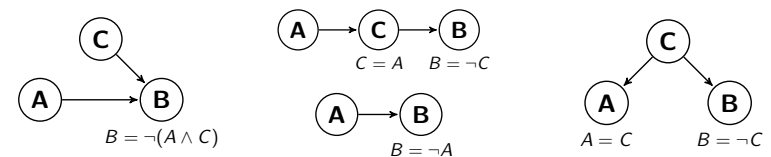
- This consolidation process gives us insight into interactions between counterfactuals
- (1) If Alice had gone to the party, Bob would have stayed home.
- (8) If Alice had gone to the party, Doug would have been home alone.
- Updating with (1) adds a covariance between A and $\neg B$ to our knowledge base
 - Alice and Bob have opposite party-attendance values
- Updating with (8) requires that A and B have the same value
- The model we build after hearing one of (1)/(8) precludes the other

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Conclusion

- We can use structured possible worlds to model dependencies, not just facts
- We propose using them to model informative counterfactuals
- Doing so gets us a natural way to typologize explanatory strategies



Conclusion

- Our analysis also neatly captures the distinction between different senses of *backtracking*
 - Classical philosophical backtrackers reverse the generally implicated direction of dependence
 - Recent psychological uses of the term refer to explanations including at least one instance of COMMON CAUSE
- Also provides insight into the mechanism that explains mutually infelicitous counterfactuals

Thank You!

Questions?

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