

Algorithmic Syntactic Causal Identification

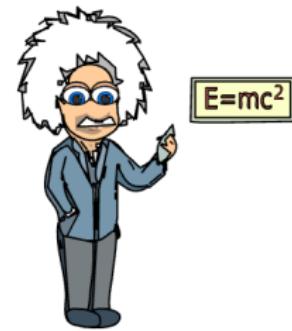
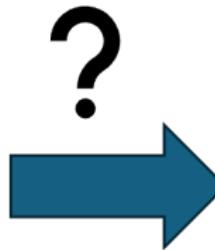
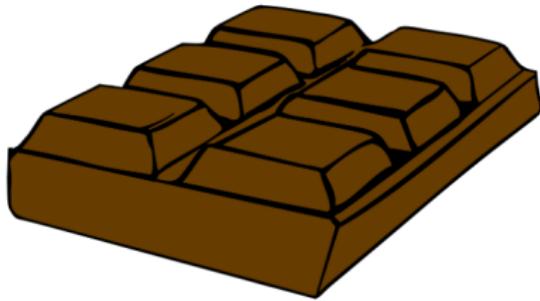
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University of Birmingham

April 14, 2025

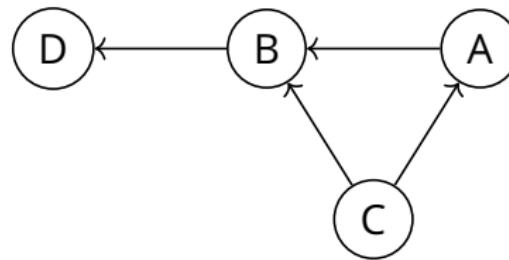
Setting the stage

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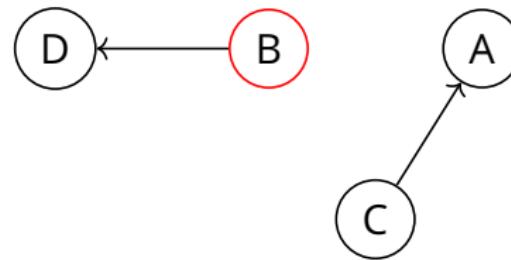
Basics of Causal inference

- Nodes represent random variables
- Directed arrows represent causal influence



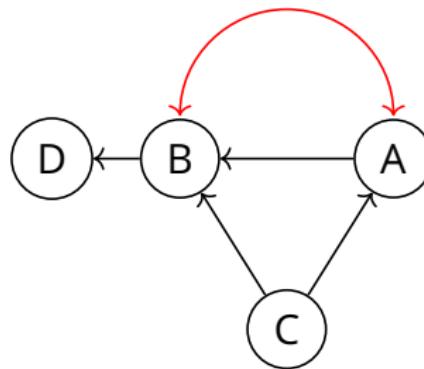
Basics of Causal inference

- Intervene by deleting parent edges
- Replace distribution with a constant distribution



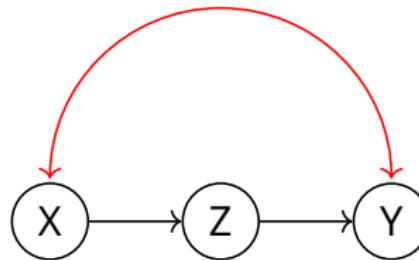
Basics of Causal inference

- In real life data is not always fully observed
- We can represent an unobserved confounder as a bidirected arc between two observed nodes



Causal identifiability

- Assess if we have enough information to answer a specific causal query
- A classic example is the *front-door* criterion



- E.g. Inferring whether or not smoking causes cancer in the lungs

Limitations of Causal inference

- How to represent Marginalization
- How to represent hidden/latent variables
- Conflation of syntax and semantics
- Causal queries outside of probability theory?

String diagrams

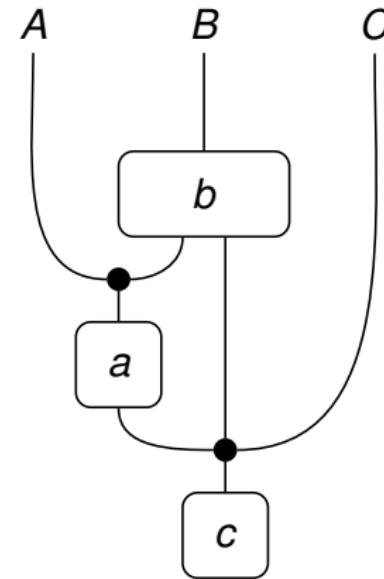
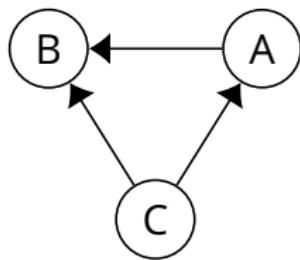


$$\begin{array}{c} a \\ \square \\ | \\ b \\ \square \\ | \\ a \\ \square \end{array} = \begin{array}{c} b \cdot a \\ \square \end{array}$$

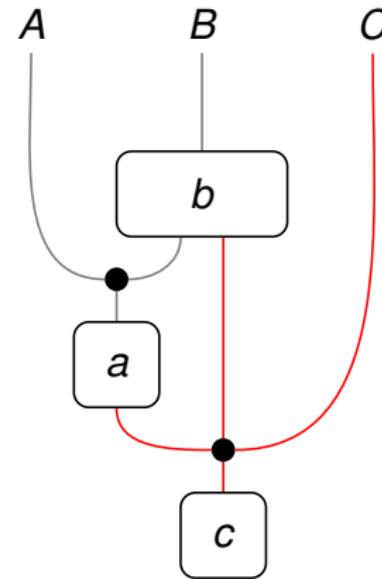
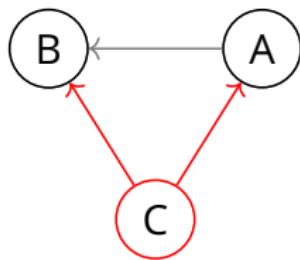
$$\begin{array}{c} a \\ \square \\ | \\ b \\ \square \\ || \\ a \otimes b \\ \square \end{array}$$

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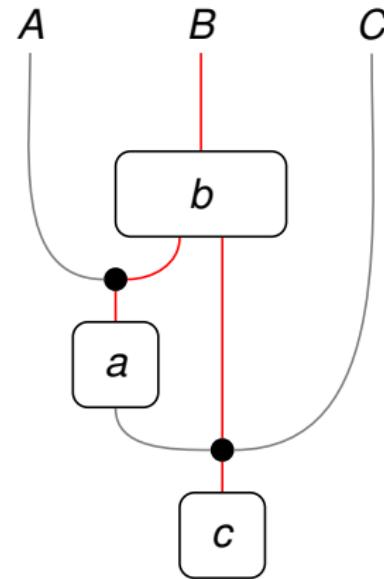
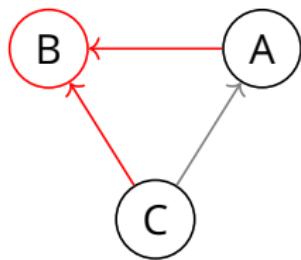
Building String Diagrams from DAGs



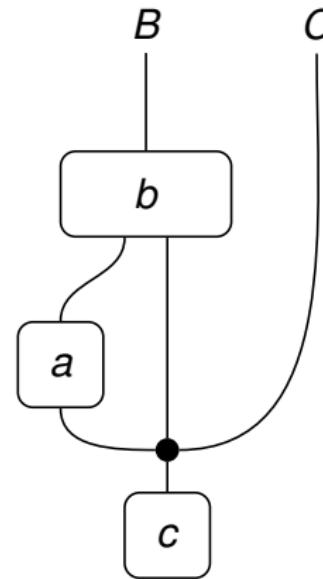
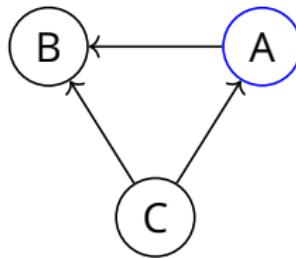
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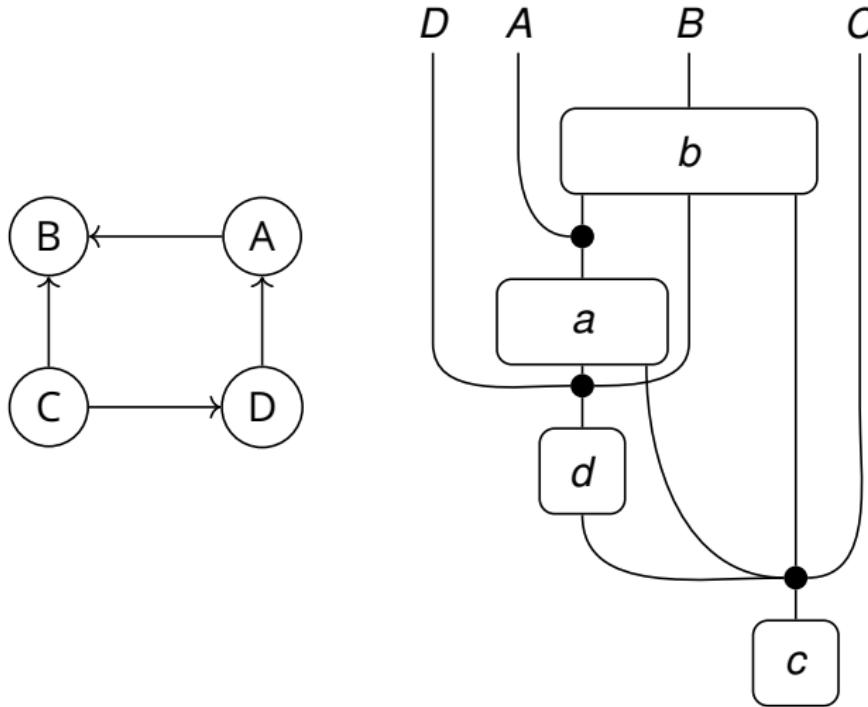
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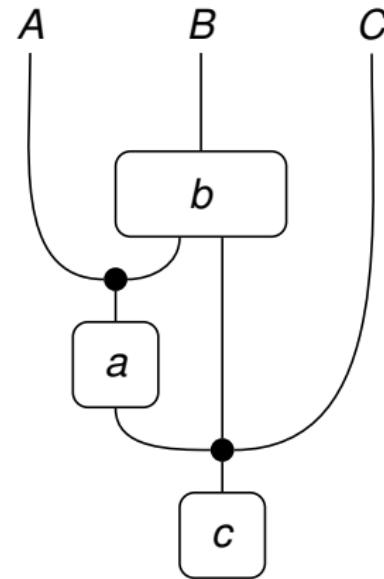
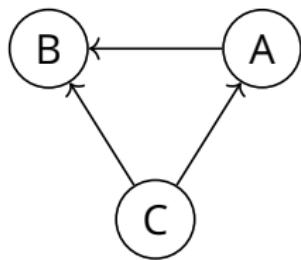
Representing unobserved variables



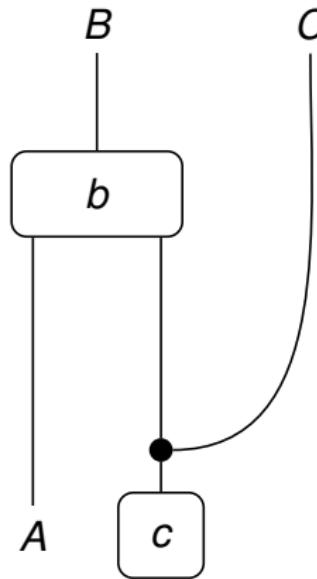
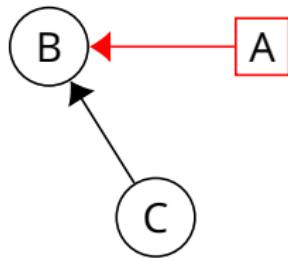
Factorized string diagrams



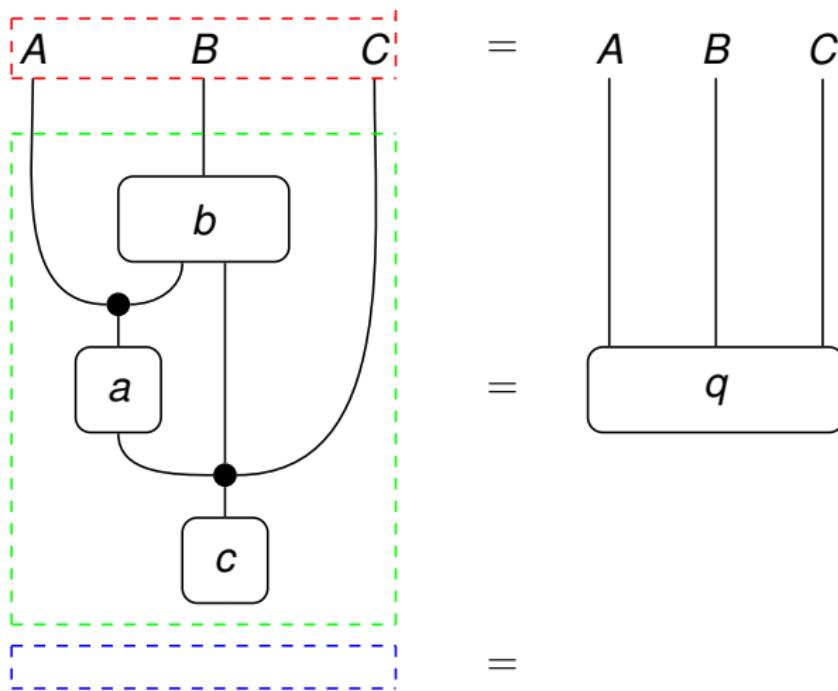
The fixing operation



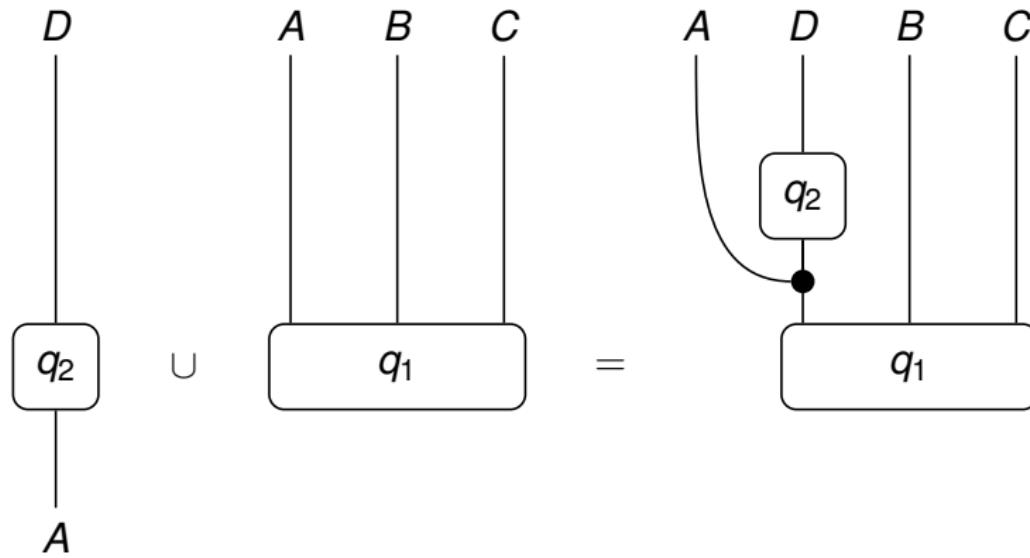
The Fixing operation



Exterior morphisms



Combining Operation



Causal Identifiability in a Syntactic setting

$$\Sigma_{\mathbf{Y}|\text{do}(\mathbf{A})}^{\mathcal{G}} = \text{Hide}_{\mathbf{Y}^* \setminus \mathbf{Y}} \left(\bigcup_{\mathbf{D}' \in \mathbf{D}^*} \text{Simple} \left(\text{Fixseq}_{\mathbf{V}^{\mathcal{G}} \setminus \mathbf{D}'} (\Sigma^{\mathcal{F}}) \right) \right)$$

Casual Identifiability in a Syntactic setting

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The factorized signature

Causal Identifiability in a Syntactic setting

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The valid sequence of applications of the fixing operation to be applied to the signature

Causal Identifiability in a Syntactic setting

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A purely algebraic formalism of the naturality of the delete operation

Causal Identifiability in a Syntactic setting

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Combining all the fixed and simplified signatures together, for each identified district

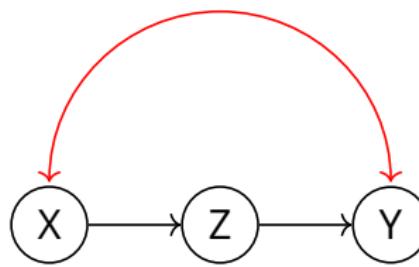
Causal Identifiability in a Syntactic setting

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Applying the deletion operation on a set of morphisms in the final combined signature

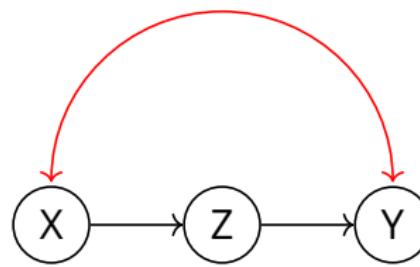
Front-door Adjustment

- Start with front-door DAG and then identify valid fixing sequences



Front-door Adjustment

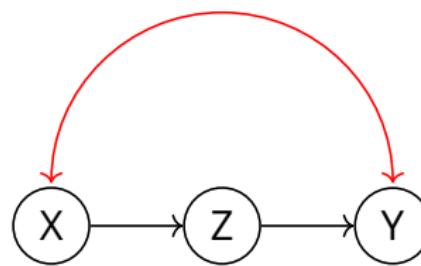
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- In this example we have;

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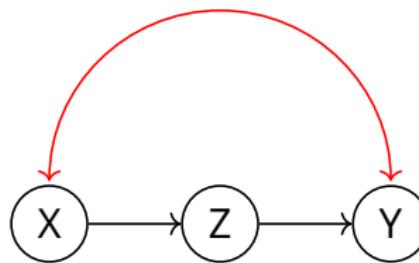
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- $\text{Fixseq}_{\{X,Z\}} = \text{Hide}_X \circ \text{Fix}_Z$

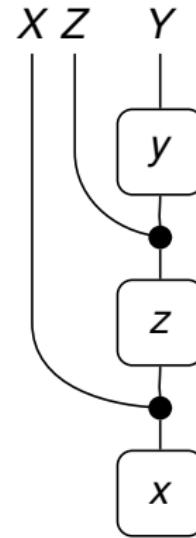
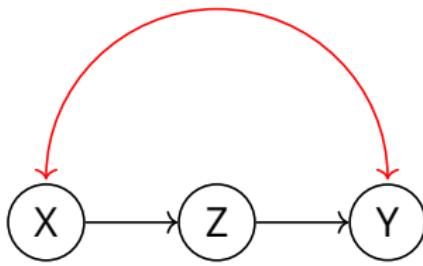
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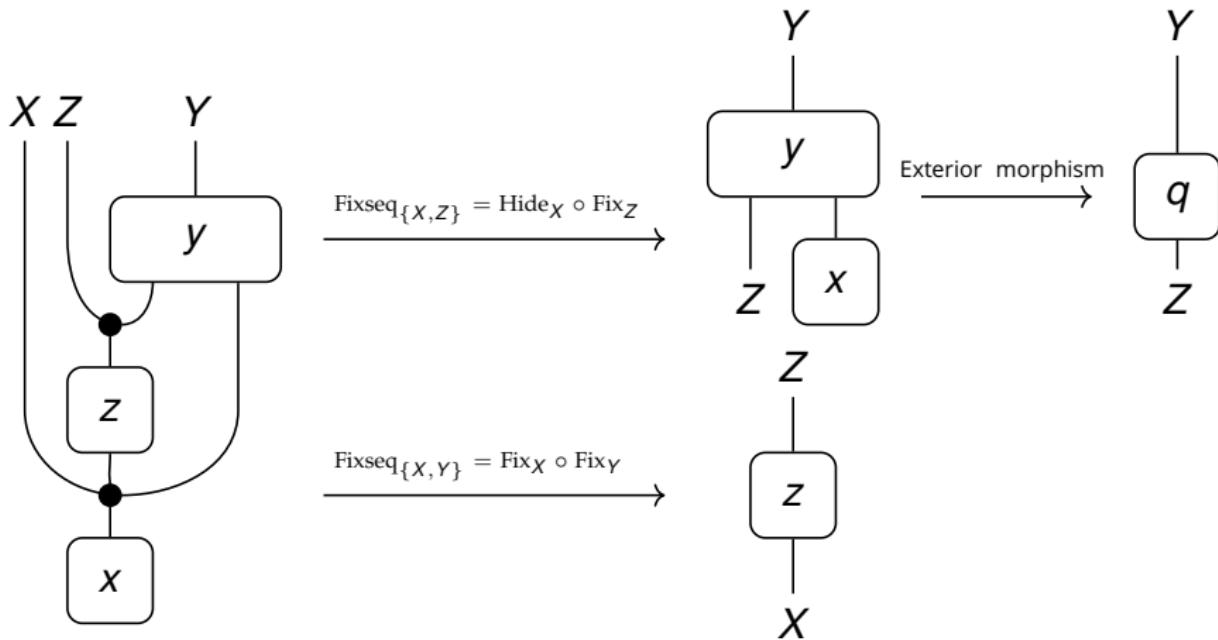


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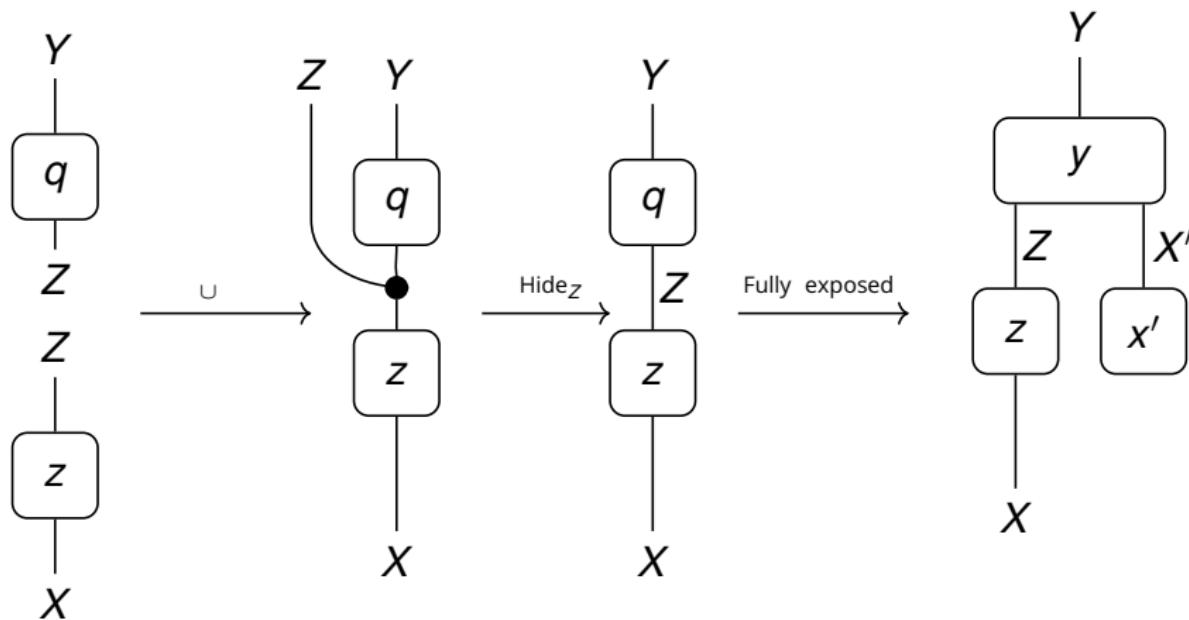
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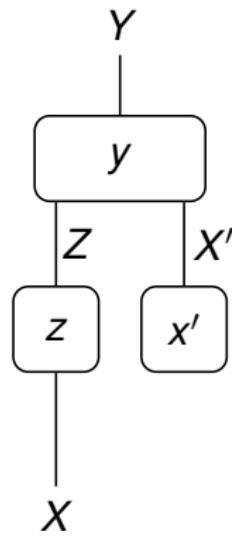
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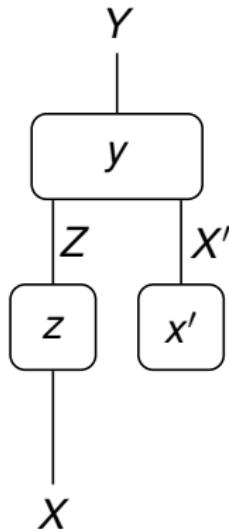
Front-door Adjustment



Interpretations

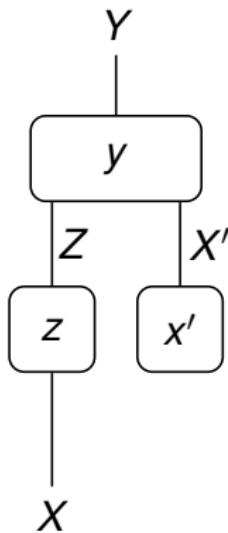


Interpretations



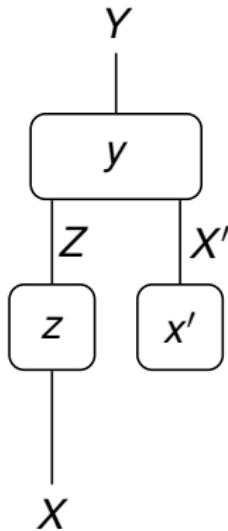
- $p(Y = y | \text{do}(X = x)) = \sum_{z \in \Omega_Z} p(Z = z | X = x) \sum_{x' \in \Omega_X} p(Y = y | X' = x', Z = z) p(X' = x')$

Interpretations



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- $f(y | \text{do}(x)) = \min_{z \in Z} f(z|x) + \min_{x' \in X} [f(y|x', z) + f(x'|)]$.

Interpretations



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- $f(y | \text{do}(x)) = \min_{z \in Z} f(z|x) + \min_{x' \in X} [f(y|x', z) + f(x')]$.
- $\text{do}(x) Ry \text{ iff } \exists z, x' \in Z, X : (xR_3 z) \wedge \left(((x', z) R_2 y) \wedge (R_3 x') \right)$

Conclusion

- We have introduced a purely syntactic form of causal identification
- This allows for causal identification in generic settings
- Interpretations can be implemented for non-probabilistic models