

Causal Inference

ICS 295 (Spring 2023, Rina Dechter)

HOMEWORK 2

Due: Saturday, February 11th, 2023

Notice that Problems 4 has a (*). This means that you are encouraged to do this question, but I will not grade this closely.

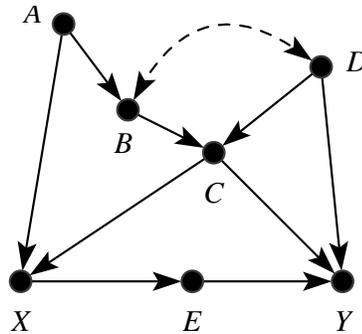
Problem 1. Modeling [5 points]

Consider the recent study of the connection between sleep quality and dementia presented here and discussed in class. <https://www.nytimes.com/2021/04/20/health/sleep-dementia-risk.html?referringSource=articleShare>.

- (a) [2 points] Provide a structural causal diagram based on your understanding of the study and assumptions made.
- (b) [1 point] Discuss the suitability of the different conclusions proposed by the study. You can focus on one or two statements.
- (c) [2 points] The article talks about associations rather than causation. In your opinion, can they claim causation and under what assumptions.

Problem 2. Understanding the Model's Granularity [10 points]

Consider the causal diagram G below.



- (a) [1 point] Determine whether the causal effect $P(y | do(x))$ is identifiable from G and $P(\mathbf{V})$, where \mathbf{V} is the set of endogenous variables. If so, show how; otherwise, provide a counter-example.
- (b) [2 points] Write an SCM \mathcal{M} that induces G and a probability distribution $P(\mathbf{V})$, with $P(\mathbf{v}) > 0$ for every \mathbf{v} . You don't need to show $P(\mathbf{V})$ in your answer.

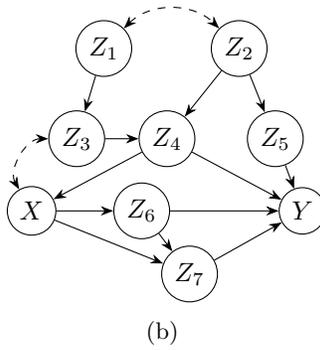
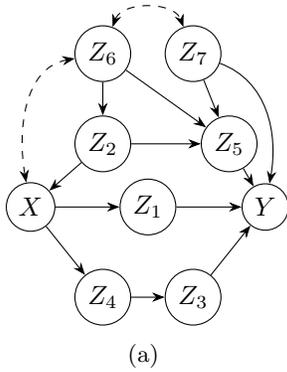
Suppose that the same system (represented by the SCM) is investigated in another study. However, in this case, only the variables $\mathbf{V}' = \{X, Y, B, C\}$ are measured.

- (c) [3 points] Write a new SCM $\mathcal{M}' = \langle \mathbf{V}', \mathbf{U}', \mathcal{F}', P(\mathbf{u}') \rangle$ corresponding to this different cut of reality, consistent with your answer to the previous question (i.e., departing from SCM \mathcal{M} written in (b)).
- (d) [1 point] Draw the causal diagram \mathcal{G}' induced by \mathcal{M}' .
- (e) [3 points] Is the effect $P(y | do(x))$ identifiable from $P(\mathbf{V}')$ and \mathcal{G}' ? Is there a back-door or front-door adjustment? Can it be solved with do-calculus?

Problem 3. Optimal Experiment Design [10 points]

An advertisement company is trying to identify the effect of a new campaign X on the click through rate Y . They

have two hypotheses about how the strategy relates to a possibly measured set of covariates \mathbf{Z} . The hypotheses are represented in the causal diagrams (a) and (b) shown below:



Variable	Cost
X	2
Y	1
Z_1	4
Z_2	2
Z_3	4
Z_4	5
Z_5	5
Z_6	2
Z_7	1

(c)

- (a) [4 points] If it exists, find a minimal admissible set for adjustment in each of the graphs.
- (b) [6 points] The company wants to minimize the measurement cost for identifying $P(y | do(x))$. Find the minimum cost ID expression based on the table (c) and justify your answer.

Problem 4. Back-door Adjustment as a Substitute for the Direct Parents [1 point]

The causal effect of the intervention $do(X = x)$ on a variable Y can be identified if all parents of X are observed and is given by

$$P(y | do(x)) = \sum_{pa_X} P(y | x, pa_X)P(pa_X). \tag{1}$$

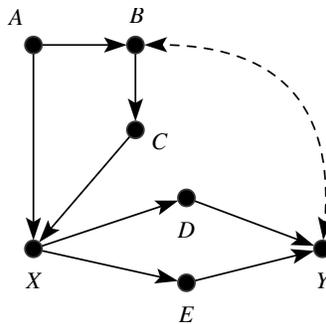
Based on this result, prove that if a set \mathbf{Z} satisfies the back-door criterion relative to X and Y in the graph \mathcal{G} , it follows that

$$P(y | do(x)) = \sum_{\mathbf{z}} P(y | x, \mathbf{z})P(\mathbf{z}). \tag{2}$$

This question is asking you to leverage Eq. (1) to prove the backdoor identification formula in Eq. (2).

Problem 5. Many Paths Lead to ID [10 points]

Consider the following causal diagram.



Give **three** different functions of the observational distribution $P(\mathbf{V})$ that are equal to the effect $P(y | do(x))$. At least one answer should correspond to a front-door case and one to a back-door case. Justify each one of the expression showing its do-calculus derivation.