

# Causal and Probabilistic Graphical Models

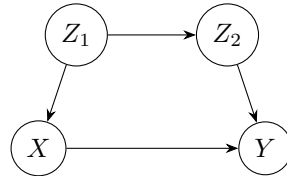
## ICS 276 (Winter 2024)

### HOMEWORK 3

Due: Monday, February 26, 2024

**Problem 1. Query Estimation [10 points]**

Consider the following graphical model:

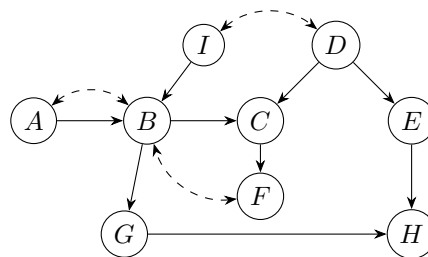


The target query is  $Q = \sum_{z_1} P(y | x, z_1)P(z_1)$ .

- (a) [5 points] Is  $Q = P(y | x)$ ? Justify your answer.
- (b) [5 points] Suppose that only  $P(X, Y, Z_2)$  is given as input. Is  $Q$  estimable? If so, show how to do it. Otherwise, explain why is that the case.

**Problem 2. d-Separation [16 points]**

- (a) Consider the following causal diagram:



For each case find a set that, when conditioned, on d-separates the given pair of (sets) variables

- (i) [4 points]  $A$  and  $F$
- (ii) [4 points]  $A$  and  $C$
- (iii) [4 points]  $D$  and  $\{F, H\}$
- (iv) [4 points]  $I$  and  $H$

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

**Problem 3. Modeling [5 points]**

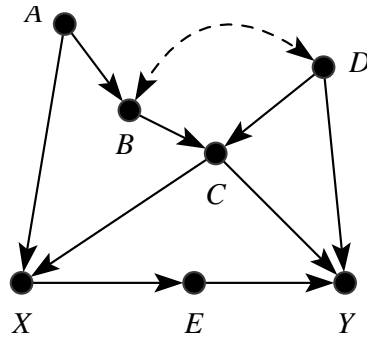
Consider the recent study of the connection between sleep quality and dementia presented here and discussed in class. <https://www.nytimes.com/2021/04/20/health/sleep-dementia-risk.html?referringSource=articleShare>.

- (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 4. Understanding the Model's Granularity [10 points]**

Consider the causal diagram  $G$  below.



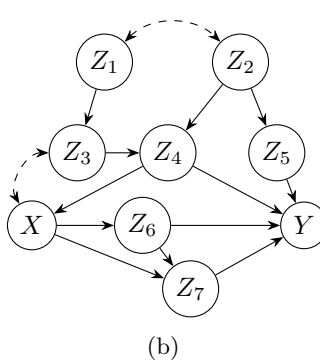
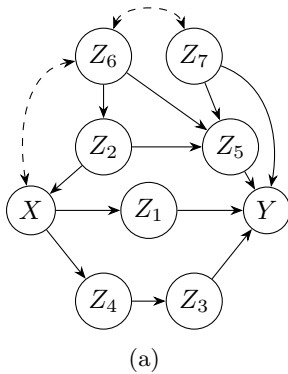
- (a) [1 point] Determine whether the causal effect  $P(y | do(x))$  is identifiable from  $G$  and  $P(\mathbf{V})$ , where  $\mathbf{V}$  is the set of endogenous variables. If so, show how; otherwise, provide a counter-example.
- (b) [2 points] Write an SCM that induces  $G$  and a probability distribution  $P(\mathbf{V})$ , 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  $M' = \langle \mathbf{V}', \mathbf{U}', \mathcal{F}', P(\mathbf{u}') \rangle$  corresponding to this different cut of reality, consistent with your answer to the previous question (i.e., departing from SCM written in (b)).
- (d) [1 point] Draw the causal diagram  $G'$  induced by the SCM  $M'$ .
- (e) [3 points] Is the effect  $P(y | do(x))$  identifiable from  $P(\mathbf{V}')$  and  $G'$ ? Is there a back-door or front-door adjustment? Can it be solved with do-calculus?

**Problem 5. 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 backdoor 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 6. 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

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

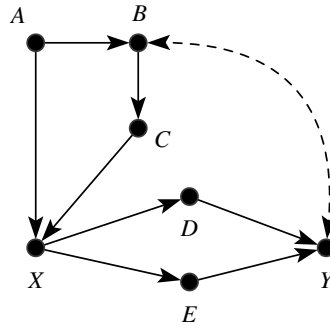
Based on this result, prove that if a set  $\mathbf{Z}$  satisfies the back-door criterion relative to  $X$  and  $Y$  in the graph, it follows that

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

This question is asking you to leverage eq3.1 to prove the backdoor identification formula in eq3.2.

**Problem 7. 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.