

6.S091 Problem Set 1

Suyeol Yun
6.S091: Causality

January 23, 2023

Preliminaries

(a)

$$\begin{aligned}\mathbb{P}_X(U, A, M, Y) &= P(U)P(A|U)P(M|A)P(Y|M, U) \\ &= \text{Ber}(0.5) \text{Ber}(U/4) \text{Ber}(0.5 + 0.1A) \text{Ber}(M/2 + U/4)\end{aligned}$$

(b)

$$\begin{aligned}\mathbb{P}_X(Y = 1) &= 0.38125 \\ \mathbb{P}_X(Y = 1 \mid M = 0, A = 0) &= 0.10714285714285714 \\ \mathbb{P}_X(Y = 1 \mid M = 0, A = 1) &= 0.25\end{aligned}$$

* I computed the above values by computing the distribution $\mathbb{P}_X(U, A, M, Y)$ as dictionary in python and marginalized. See https://github.com/syyunn/6.S091/blob/main/pset1/problem_1b.py

Interventional

(c)

$$\begin{aligned}\mathbb{P}_X(U, A = 1, M, Y \mid \text{do}(A = 1)) &= P(U)P(A = 1|A = 1)P(M|A = 1)P(Y|M, U, A = 1) \\ &= \text{Ber}(0.5) \cdot 1 \cdot \text{Ber}(0.6) \cdot \text{Ber}(M/2 + U/4) \\ &= \text{Ber}(0.5) \cdot \text{Ber}(0.6) \cdot \text{Ber}(M/2 + U/4)\end{aligned}$$

In the similar vein, we can compute

$$\begin{aligned}\mathbb{P}_X(U, A = 1, M, Y \mid \text{do}(A = 0)) &= P(U)P(A = 0|A = 0)P(M|A = 0)P(Y|M, U, A = 1) \\ &= \text{Ber}(0.5) \cdot 1 \cdot \text{Ber}(0.5) \cdot \text{Ber}(M/2 + U/4) \\ &= \text{Ber}(0.5) \cdot \text{Ber}(0.5) \cdot \text{Ber}(M/2 + U/4)\end{aligned}$$

(d)

$$\mathbb{P}_X(Y \mid \text{do}(A = 1)) = \begin{cases} 0.425 & \text{if } Y = 1 \\ 0.575 & \text{if } Y = 0 \end{cases} \quad (1)$$

$$\mathbb{P}_X(Y \mid \text{do}(A = 0)) = \begin{cases} 0.375 & \text{if } Y = 1 \\ 0.625 & \text{if } Y = 0 \end{cases} \quad (2)$$

* I computed the above values by computing the distribution $\mathbb{P}_X(U, A, M, Y)$ as dictionary in python and marginalized. See https://github.com/syyunn/6.S091/blob/main/pset1/problem_1d.doA0.py and https://github.com/syyunn/6.S091/blob/main/pset1/problem_1d.doA1.py

(e)

In case of $do(A = 0)$, we have

$$\begin{aligned}\mathbb{P}_X(Y = 1 \mid do(A = 0)) &= 0.375 \quad \text{from (d)} \\ \sum_{u \in \{0,1\}} \mathbb{P}_X(Y = 1 \mid A = 0, U = u) \mathbb{P}_X(U = u) &= \mathbb{P}_X(Y = 1 \mid A = 0, U = 1) \mathbb{P}_X(U = 1) + \mathbb{P}_X(Y = 1 \mid A = 0, U = 0) \mathbb{P}_X(U = 0) \\ &= 0.5 * 1/2 + 0.25 * 1/2 \\ &= 0.375 \quad (\text{match})\end{aligned}$$

$$\begin{aligned}\mathbb{P}_X(Y = 0 \mid do(A = 0)) &= 0.625 \quad \text{from (d)} \\ \sum_{u \in \{0,1\}} \mathbb{P}_X(Y = 0 \mid A = 0, U = u) \mathbb{P}_X(U = u) &= \mathbb{P}_X(Y = 0 \mid A = 0, U = 1) \mathbb{P}_X(U = 1) + \mathbb{P}_X(Y = 0 \mid A = 0, U = 0) \mathbb{P}_X(U = 0) \\ &= 0.5 * 1/2 + 0.75 * 1/2 \\ &= 0.625 \quad (\text{match})\end{aligned}$$

In case of $do(A = 1)$, we have

$$\begin{aligned}\mathbb{P}_X(Y = 1 \mid do(A = 1)) &= 0.425 \quad \text{from (d)} \\ \sum_{u \in \{0,1\}} \mathbb{P}_X(Y = 1 \mid A = 1, U = u) \mathbb{P}_X(U = u) &= \mathbb{P}_X(Y = 1 \mid A = 1, U = 1) \mathbb{P}_X(U = 1) + \mathbb{P}_X(Y = 1 \mid A = 1, U = 0) \mathbb{P}_X(U = 0) \\ &= 0.55 * 1/2 + \text{undefined} \\ &= \text{undefined}\end{aligned}$$

$$\begin{aligned}\mathbb{P}_X(Y = 0 \mid do(A = 1)) &= 0.625 \quad \text{from (d)} \\ \sum_{u \in \{0,1\}} \mathbb{P}_X(Y = 0 \mid A = 1, U = u) \mathbb{P}_X(U = u) &= \mathbb{P}_X(Y = 0 \mid A = 1, U = 1) \mathbb{P}_X(U = 1) + \mathbb{P}_X(Y = 0 \mid A = 1, U = 0) \mathbb{P}_X(U = 0) \\ &= 0.45 * 1/2 + \text{undefined} \\ &= \text{undefined}\end{aligned}$$

(f)

In case of $do(A = 0)$, we have

$$\begin{aligned}\mathbb{P}_X(Y = 1 \mid do(A = 0)) &= 0.375 \quad \text{from (d)} \\ \sum_{m \in \{0,1\}} \mathbb{P}_X(M = m \mid A = 0) \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 1 \mid M = m, A = a') \mathbb{P}_X(A = a')) \\ &= \mathbb{P}_X(M = 0 \mid A = 0) \cdot \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 1 \mid M = 0, A = a') \mathbb{P}_X(A = a')) \\ &\quad + \mathbb{P}_X(M = 1 \mid A = 0) \cdot \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 1 \mid M = 1, A = a') \mathbb{P}_X(A = a')) \\ &= 1/2 \cdot (0.10714285714285714 \cdot 0.875 + 0.25 \cdot 0.125) + 1/2 \cdot (0.6071428571428571 \cdot 0.875 + 0.75 \cdot 0.125) \\ &= 0.375 \quad (\text{match}) \\ \mathbb{P}_X(Y = 0 \mid do(A = 0)) &= 0.625 \quad \text{from (c)} \\ \sum_{m \in \{0,1\}} \mathbb{P}_X(M = m \mid A = 0) \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 0 \mid M = m, A = a') \mathbb{P}_X(A = a')) \\ &= \mathbb{P}_X(M = 0 \mid A = 0) \cdot \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 0 \mid M = 0, A = a') \mathbb{P}_X(A = a')) \\ &\quad + \mathbb{P}_X(M = 1 \mid A = 0) \cdot \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 0 \mid M = 1, A = a') \mathbb{P}_X(A = a')) \\ &= 1/2 \cdot (0.8928571428571429 \cdot 0.875 + 0.7500000000000001 \cdot 0.125) + 1/2 \cdot (0.39285714285714285 \cdot 0.875 + 0.25 \cdot 0.125) \\ &= 0.625000000000000218750 \quad (\text{match})\end{aligned}$$

In case of $do(A = 1)$, we have

$$\mathbb{P}_X(Y = 1 \mid do(A = 1)) = 0.425 \quad \text{from (d)}$$

$$\begin{aligned} & \sum_{m \in \{0,1\}} \mathbb{P}_X(M = m \mid A = 1) \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 1 \mid M = m, A = a') \mathbb{P}_X(A = a')) \\ &= \mathbb{P}_X(M = 0 \mid A = 1) \cdot \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 1 \mid M = 0, A = a') \mathbb{P}_X(A = a')) \\ &+ \mathbb{P}_X(M = 1 \mid A = 1) \cdot \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 1 \mid M = 1, A = a') \mathbb{P}_X(A = a')) \\ &= 0.4 \cdot (0.10714285714285714 \cdot 0.875 + 0.25 \cdot 0.125) + 0.6 \cdot (0.6071428571428571 \cdot 0.875 + 0.75 \cdot 0.125) \\ &= 0.42499999999999997650 \quad (\text{match}) \end{aligned}$$

$$\mathbb{P}_X(Y = 0 \mid do(A = 1)) = 0.575 \quad \text{from (c)}$$

$$\begin{aligned} & \sum_{m \in \{0,1\}} \mathbb{P}_X(M = m \mid A = 1) \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 0 \mid M = m, A = a') \mathbb{P}_X(A = a')) \\ &= \mathbb{P}_X(M = 0 \mid A = 1) \cdot \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 0 \mid M = 0, A = a') \mathbb{P}_X(A = a')) \\ &+ \mathbb{P}_X(M = 1 \mid A = 1) \cdot \sum_{a' \in \{0,1\}} (\mathbb{P}_X(Y = 0 \mid M = 1, A = a') \mathbb{P}_X(A = a')) \\ &= 0.4 \cdot (0.8928571428571429 \cdot 0.875 + 0.7500000000000001 \cdot 0.125) + 0.6 \cdot (0.39285714285714285 \cdot 0.875 + 0.25 \cdot 0.125) \\ &= 0.57500000000000001625 \quad (\text{match}) \end{aligned}$$

Problem 2

(a)

Let's take an arbitrary node $\gamma_m \in \gamma$. Then γ_m should be unblocked since γ is a d-connecting path and it means all nodes in γ are unblocked. Then since γ_m is unblocked, we can think of two cases where γ_m is a collider or γ_m is a non-collider. First, in case of γ_m is a collider, then $\gamma_m \in S$ or its descendant $de_{\mathcal{G}}(\gamma_m) \in S$. If $\gamma_m \in S$, $\gamma_m \in V$ since V includes S itself. If $de_{\mathcal{G}}(\gamma_m) \in S$, $\gamma_m \in an_{\mathcal{G}}(S)$, and it implies $\gamma_m \in V$. Second, in case of γ_m is a non-collider, $\gamma_m \notin S$. In this case, let's think about the sub-path $\gamma_m \rightarrow \gamma_{m+1}, \gamma_{m+2}, \dots, b$ where b is the destination of the path γ . Among the nodes in this sub-path, if there exists a node that is a collider, then this node is in S or an ancestor of S as we've shown in the first scenario. Then γ_m becomes an ancestor of S and it implies $\gamma_m \in V$. In other case, if there is no collider in the sub-path, then γ_m is an ancestor of b and it implies $\gamma_m \in V$. Therefore, in any cases, $\gamma_m \in V$.

(b)

From (a), we know that $\gamma_m \in S$ only in case that γ_m is a collider. By the definition of moral graph, if γ_m is a collider, we can marry γ_{m-1} and γ_{m+1} which are left and right nodes of γ_m . Therefore, with given path γ , we can construct a new path γ' by visiting each node in γ and marrying left and right nodes when the node we're visiting is a collider. Then this new path γ' doesn't pass through S and γ' is a valid path in moral graph $\overline{\mathcal{G}[V]}$ since its all nodes and edges are included in $\overline{\mathcal{G}[V]}$.

Problem 3

(a)

$$\begin{aligned} \mathbb{P}_X(Y = 1 \mid A = N) &= \sum_{S \in \{L,R\}} \mathbb{P}_X(Y = 1 \mid S, A = N) \mathbb{P}_X(S) \\ &= 0.73 \cdot 0.49 + 0.93 \cdot 0.51 \\ &= 0.832 \\ \mathbb{P}_X(Y = 1 \mid A = O) &= \sum_{S \in \{L,R\}} \mathbb{P}_X(Y = 1 \mid S, A = O) \mathbb{P}_X(S) \\ &= 0.69 \cdot 0.49 + 0.87 \cdot 0.51 \\ &= 0.7818 \end{aligned}$$

(b)

Since S satisfies backdoor criterion, $\mathbb{P}_{\mathcal{X}}(Y = 1 \mid \text{do}(A = N)) = \sum_{\mathbf{s}} \mathbb{P}_{\mathcal{X}}(Y \mid A = N, \mathbf{S} = \mathbf{s}) \mathbb{P}_{\mathcal{X}}(\mathbf{S} = \mathbf{s}) = 0.832$ and $\mathbb{P}_{\mathcal{X}}(Y = 1 \mid \text{do}(A = O)) = \sum_{\mathbf{s}} \mathbb{P}_{\mathcal{X}}(Y \mid A = O, \mathbf{S} = \mathbf{s}) \mathbb{P}_{\mathcal{X}}(\mathbf{S} = \mathbf{s}) = 0.7818$.

Problem 4

β_{aw}	$\hat{\beta}_{aw}$	$\hat{\beta}_{yw}$	$\widehat{\beta}_{yw}/\widehat{\beta}_{aw}$	$ \mathbb{E}[Y A] - \widehat{\beta}_{yw}/\widehat{\beta}_{aw} $
5	4.52	33.84	7.47	0.03
0.5	0.72	5.54	7.60	0.1
0.05	0.32	2.49	7.62	0.12

As shown in the table, the ratio well estimates the true $\mathbb{E}[Y|A] = 0.75$. Also we can notice that the absolute value of bias $|\mathbb{E}[Y|A] - \widehat{\beta}_{yw}/\widehat{\beta}_{aw}|$ increases as β_{aw} decreases.

* I computed the above values by computing the distribution $\mathbb{P}_{\mathcal{X}}(U, A, M, Y)$ as dictionary in python and marginalized. See <https://github.com/syyunn/6.S091/blob/main/pset1/problem4.py>