

CS 475 Machine Learning: Homework 6

Graphical Models 2

Due: Friday December 4, 2015, 11:59pm

50 Points Total

Version 1.0

Li-Yi Lin/lilin34@jhu.edu

1 Analytical (20 points)

1. (10 points) Causal Effect Identification Use do-calculus to give an identifying expression for $p(x_5 | do(x_3))$ in terms of $p(x_1, x_2, x_3, x_4, x_5)$ in a hidden variable causal model in Fig. 1. Note: your expression is not allowed to refer to h_1, h_2, h_3 as those variables are not observed.

Ans:

$$\begin{aligned}
& P(X_5 | do(x_3)) \\
& \stackrel{p}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} P(X_5, x_1, x_2, x_4 | do(x_3)) \\
& \stackrel{c}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} P(X_5 | x_1, x_2, x_4, do(x_3)) P(x_1, x_2, x_4 | do(x_3)) \\
& \stackrel{c}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} P(X_5 | x_1, x_2, x_4, do(x_3)) P(x_4 | x_1, x_2, do(x_3)) P(x_1, x_2 | do(x_3)) \\
& \stackrel{c}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} P(X_5 | x_1, x_2, x_4, do(x_3)) P(x_4 | x_1, x_2, do(x_3)) P(x_2 | x_1, do(x_3)) P(x_1 | do(x_3)) \\
& \stackrel{3}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} P(X_5 | x_1, x_2, x_4, do(x_3)) P(x_4 | x_1, x_2, do(x_3)) P(x_2 | x_1, do(x_3)) P(x_1) \\
& \stackrel{2}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} P(X_5 | x_1, x_2, x_4, do(x_3)) P(x_4 | x_1, x_2, x_3) P(x_2 | x_1, do(x_3)) P(x_1) \\
& \stackrel{2}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} P(X_5 | x_1, x_2, do(x_4), do(x_3)) P(x_4 | x_1, x_2, x_3) P(x_2 | x_1, do(x_3)) P(x_1) \\
& \stackrel{3}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} P(X_5 | x_1, x_2, do(x_4)) P(x_4 | x_1, x_2, x_3) P(x_2 | x_1, do(x_3)) P(x_1) \\
& \stackrel{\text{add } x'_3}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} \sum_{x'_3} P(x_5, x'_3 | x_1, x_2, do(x_4)) P(x_4 | x_1, x_2, x_3) P(x_2 | x_1, do(x_3)) P(x_1) \\
& \stackrel{c}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} \sum_{x'_3} P(X_5 | x'_3, x_1, x_2, do(x_4)) P(x'_3 | x_1, x_2, do(x_4)) P(x_4 | x_1, x_2, x_3) P(x_2 | x_1, do(x_3)) P(x_1) \\
& \stackrel{2}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} \sum_{x'_3} P(X_5 | x'_3, x_1, x_2, x_4) P(x'_3 | x_1, x_2, do(x_4)) P(x_4 | x_1, x_2, x_3) P(x_2 | x_1, do(x_3)) P(x_1) \\
& \stackrel{3}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} \sum_{x'_3} P(X_5 | x'_3, x_1, x_2, x_4) P(x'_3 | x_1, x_2) P(x_4 | x_1, x_2, x_3) P(x_2 | x_1, do(x_3)) P(x_1) \\
& \stackrel{3}{=} \sum_{x_1} \sum_{x_2} \sum_{x_4} \sum_{x'_3} P(X_5 | x'_3, x_1, x_2, x_4) P(x'_3 | x_1, x_2) P(x_4 | x_1, x_2, x_3) P(x_2 | x_1) P(x_1)
\end{aligned}$$

2. (10 points) Structure Learning We have data on four variables x_1, x_2, x_3, x_4 , and running a set of hypothesis tests we learned that the following set of conditional independences hold in our data: $(x_2 \perp\!\!\!\perp x_3 \mid x_1)$, $(x_4 \perp\!\!\!\perp x_1 \mid x_2, x_3)$. Assuming our data was generated by a DAG model (Bayesian network), give the set of DAGs consistent with

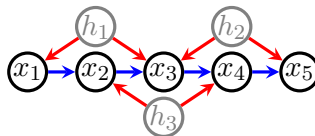
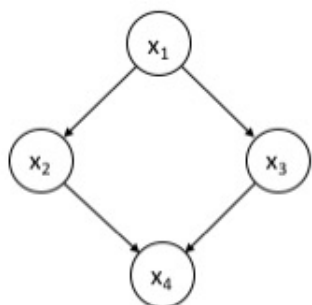


Figure 1: A hidden variable causal DAG.

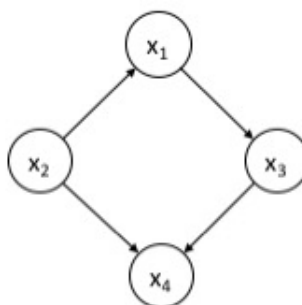
the set of independences we see. Are there any causal effects with a single outcome (e.g. $p(x_i \mid \text{do}(\mathbf{x}))$, where $\mathbf{x} = \{x_j \mid i \neq j\}$) that could be identified by the same expression, *regardless* of which DAG in the set is the true DAG?

Ans:

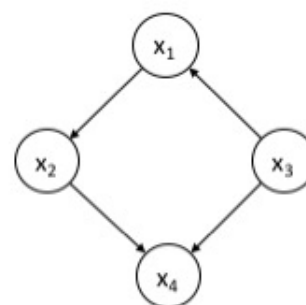
The corresponding DAGs are shown below.



DAG 1



DAG 2



DAG 3

The three DAG graphs above have the same skeleton and the same unshielded collider that is node X_4 . Therefore, regardless which one is the real graph, they will result in a single outcome by the theorem "Verma and Pearl". They are observational equivalence