$\begin{array}{c} \text{Machine Learning 2 - Homework} \\ \text{Week 6} \end{array}$

Minh Ngo
MSc Artificial Intelligence
University of Amsterdam
nlminhtl@gmail.com

May 21, 2016

Collaborators: Riaan Zoetmulder, Arthur Bražinskas

Problem 2

a Possible structures are presented in the figure 1.

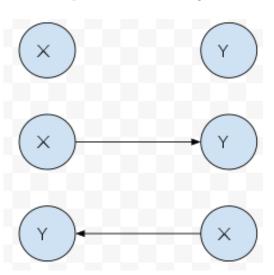


Figure 1: (i) X and Y are independent, (ii) X causes Y, (iii) Y causes X

b (i)
$$p(X, Y) = p(X)p(Y)$$

(ii)
$$p(X,Y) = p(Y|X)p(X)$$

(iii)
$$p(X,Y) = p(X|Y)p(Y)$$

c (i)
$$p(Y|X) = \frac{p(X,Y)}{p(X)} = \frac{p(X)p(Y)}{p(X)} = p(Y)$$

(ii)
$$p(Y|X) = \frac{p(X,Y)}{p(X)} = \frac{p(Y|X)p(X)}{p(X)} = p(Y|X)$$

(iii)
$$p(Y|X) = \frac{p(X,Y)}{p(X)} = \frac{p(X|Y)p(Y)}{\sum_{Y} p(X|Y)p(Y)}$$

- d (i) p(Y|do(X)) = p(Y)
 - (ii) p(Y|do(X)) = p(Y|X)
 - (iii) p(Y|do(X)) = p(Y)
- e p(Y|X) probability that somebody gets lung cancer, given the observation that the person smokes.

p(Y|do(X)) - probability that somebody gets lung cancer if we force the person to smoke [Slides, page 20].

Problem 3

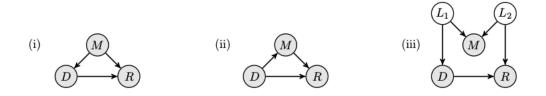


Figure 2: Different hypothetical causal models.

1a
$$p(R = 1|D = 1) = \frac{20}{40} = 0.5$$

 $p(R = 1|D = 0) = \frac{16}{40} = 0.4$

1b p(R=1|D=1) > p(R=1|D=0) therefore a new patient should take the drug.

2a
$$p(R = 1|D = 1, M = 1) = \frac{18}{30} = 0.6$$

 $p(R = 1|D = 0, M = 1) = \frac{7}{10} = 0.7$
 $p(R = 1|D = 1, M = 0) = \frac{2}{10} = 0.2$
 $p(R = 1|D = 0, M = 0) = \frac{9}{30} = 0.3$

2b Chance to recover is lower if a patient takes the drug, therefore a new patient of both gender shouldn't be recommended to take the drug.

3
$$p(R = 1|D = 1) = p(R = 1|D = 1, M = 1)p(M = 1|D = 1) + p(R = 1|D = 1, M = 0)p(M = 0|D = 1) = 0.6 * 0.75 + 0.2 * 0.25 = 0.5$$

 $p(R = 1|D = 0) = p(R = 1|D = 0, M = 1)p(M = 1|D = 0) + p(R = 1|D = 0, M = 0)p(M = 0|D = 0) = 0.7 * 0.25 + 0.3 * 0.75 = 0.4$
Since $p(R = 1|D = 1) > p(R = 1|D = 0)$ we should suggest drugs that contradicts [2b] but doesn't contradict [1b].

- 4a (1) $S = \{M\}, D, R \notin S$
 - (2) M is not a descendant of D
 - (3) M blocks all back-door path $D \leftarrow ..R$

$$\Rightarrow p(R|do(D)) = \int p(R|D,m)p(m)dm = \sum_{M} p(R|D,M)p(M)$$

4b

$$p(R|D) = \frac{\int p(R|D, M)p(D|M)p(M)dM}{\int \int p(R|D, M)p(D|M)p(M)dMdD}$$
$$= \frac{\sum_{M} p(R|D, M)p(D|M)p(M)}{\sum_{M} \sum_{D} p(R|D, M)p(D|M)p(M)}$$
(1)

Consequently $p(R|do(D)) \neq p(R|D)$.

4c

$$p(R = 1|do(D = 1)) = \sum_{M} p(R = 1|D = 1, M)p(M)$$

$$= 0.6 * 0.5 + 0.2 * 0.5 = 0.4$$
(2)

$$p(R = 1|do(D = 0)) = \sum_{M} p(R = 1|D = 0, M)p(M)$$

$$= 0.7 * 0.5 + 0.3 * 0.5 = 0.5$$
(3)

p(R=1|do(D=0)) > p(R=1|do(D=1)), therefore we shouldn't recommend drugs.

5a

$$p(R = 1|do(D)) = \sum_{M} p(R = 1|D, M)p(M|D) = p(R = 1|D)$$
 (4)

5b Yes, p(R|do(D)) = p(R|D)

5c
$$p(R = 1|do(D = 1)) = 0.6 * 0.75 + 0.2 * 0.25 = 0.5$$

 $p(R = 1|do(D = 0)) = 0.7 * 0.25 + 0.3 * 0.75 = 0.4$
 $p(R = 1|do(D = 1)) > p(R = 1|do(D = 0))$, therefore we should recommend drugs.

- 6a D drug / non drug, R recover / non recover, L2 good doctor / bad doctor, M good mood / bad mood, L1 stress / non stress.
- 6b Since $D \leftarrow L_1 \rightarrow M \leftarrow L_2 \rightarrow R$ is initially blocked p(R|do(D)) = p(R|D)

6c

$$p(R|D) = \frac{\int \int p(L_1)p(L_2)p(R|D, L_2)p(D|L_1)dL_1dL_2}{\int \int \int p(L_1)p(L_2)p(R|D, L_2)p(D|L_1)dL_1dL_2dR}$$
(5)

Consequently $p(R|do(D)) \neq p(R|D)$

6d
$$p(R = 1|do(D = 1)) = p(R = 1|D = 1) = 0.5$$

 $p(R = 1|do(D = 0)) = p(R = 1|D = 0) = 0.4$
 $p(R = 1|do(D = 1)) > p(R = 1|do(D = 0))$, therefore drug should be recommended.

Problem 5

1

$$p(Y|do(X), X_{pa(X)}) = \frac{p(Y, X_{pa(X)}|do(X))}{p(X_{pa(X)}|do(X))}$$

$$= \frac{p(Y|X, X_{pa(X)})p(X_{pa(X)})}{p(X_{pa(X)})}$$

$$= p(Y|X, X_{pa(X)}) \quad q.e.d.$$
(6)

$$p(X_{pa(X)}|do(X)) = p(X_{pa(X)}) \tag{7}$$

Since they are independent. Taking an example: a student doing a homework. His excellent grade (X) is dependent on his material understanding of each topic from the homework $(X_{pa(X)})$. If he will just copy right answers (and let's assume that it's allowed for simplicity) then he will get an excellent grade independent from his understanding.

$$p(Y|do(X)) = \int p(Y, X_{pa(X)}|do(X))dX_{pa(X)}$$

$$= \int p(Y|X_{pa(X)}|, do(X))p(X_{pa(X)}|do(X))dX_{pa(X)}$$
(8)
$$= \int p(Y|X_{pa(X)}, X)p(X_{pa(X)})dX_{pa(X)} \quad q.e.d.$$

It's valuable to mention that we can do marginalization over parents of X even if any of them are independent of Y, since integration over this variable will be equal to 1.