# Machine Learning Exercise Sheet 2

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### October 26, 2021

# Exercise 1

### (a)

Given is the DAG  $G_1$  with the following associated probabilities:

• 
$$p(A) = 0.3$$

• 
$$p(B \mid A) = 0.2$$

• 
$$p(B \mid \neg A) = 0.4$$

• 
$$p(C \mid B) = 0.7$$

• 
$$p(C \mid \neg B) = 0.5$$

From there we can deduct the following probabilities:

• 
$$p(\neg A) = 0.7$$

• 
$$p(\neg B \mid A) = 0.8$$

• 
$$p(\neg B \mid \neg A) = 0.6$$

• 
$$p(\neg C \mid B) = 0.3$$

• 
$$p(\neg C \mid \neg B) = 0.5$$

From  $G_1$  we can deduct the joint probability distribution:

$$p(A, B, C) = p(A) \cdot p(B \mid A) \cdot p(C \mid B)$$

Using this Probability distribution we can calculate P(B):

$$\begin{split} p(B) &= \sum_{A,C} p(A,B=1,C) \\ &= p(A) \cdot p(B \mid A) \cdot p(C \mid B) \\ &+ p(\neg A) \cdot p(B \mid \neg A) \cdot p(C \mid B) \\ &+ p(\neg A) \cdot p(B \mid \neg A) \cdot p(\neg C \mid B) \\ &+ p(A) \cdot p(B \mid A) \cdot p(\neg C \mid B) \\ &= 0.3 \cdot 0.2 \cdot 0.7 \\ &+ 0.7 \cdot 0.4 \cdot 0.7 \\ &+ 0.7 \cdot 0.4 \cdot 0.3 \\ &+ 0.3 \cdot 0.2 \cdot 0.3 \\ &= 0.34 \end{split}$$

and P(C):

$$\begin{split} p(C) &= \sum_{A,B} p(A,B,C=1) \\ &= p(A) \cdot p(B \mid A) \cdot p(C \mid B) \\ &+ p(\neg A) \cdot p(B \mid \neg A) \cdot p(C \mid B) \\ &+ p(\neg A) \cdot p(\neg B \mid \neg A) \cdot p(C \mid \neg B) \\ &+ p(A) \cdot p(\neg B \mid A) \cdot p(C \mid \neg B) \\ &= 0.3 \cdot 0.2 \cdot 0.7 \\ &+ 0.7 \cdot 0.4 \cdot 0.7 \\ &+ 0.7 \cdot 0.6 \cdot 0.5 \\ &+ 0.3 \cdot 0.8 \cdot 0.5 \\ &= 0.568 \end{split}$$

(b)

Given is a DAG  $\mathcal{G}_2$  and the associated probabilities:

- p(A) = 0.3
- $p(B \mid A) = 0.2$
- $p(B \mid \neg A) = 0.4$
- $p(c \mid A) = 0.7$
- $p(C \mid \neg A) = 0.6$
- $p(D \mid B, C) = 0.9$

- $p(D \mid B, \neg C) = 0.5$
- $p(D \mid \neg B, C) = 0.3$
- $p(D \mid \neg B, \neg C) = 0.3$

From there we can deduct the following probabilities:

- $p(\neg A) = 0.7$
- $p(\neg B \mid A) = 0.8$
- $p(\neg B \mid \neg A) = 0.6$
- $p(\neg c \mid A) = 0.3$
- $p(\neg C \mid \neg A) = 0.4$
- $p(\neg D \mid B, C) = 0.1$
- $p(\neg D \mid B, \neg C) = 0.5$
- $p(\neg D \mid \neg B, C) = 0.7$
- $p(\neg D \mid \neg B, \neg C) = 0.7$

From the  $G_2$  we can deduct the following joint Probability distribution:

$$p(A, B, C, D) = p(A) \cdot p(C \mid A) \cdot p(B \mid A) \cdot p(D \mid B, C)$$

Using the joint Probability distribution and the probabilities from above, we can calculate p(B):

$$\begin{split} p(B) &= \sum_{A,C,D} p(A,B=1,C,D) \\ &= p(A) \cdot p(C \mid A) \cdot p(B \mid A) \cdot p(D \mid B,C) \\ &+ p(A) \cdot p(C \mid A) \cdot p(B \mid A) \cdot p(\neg D \mid B,C) \\ &+ p(A) \cdot p(\neg C \mid A) \cdot p(B \mid A) \cdot p(D \mid B, \neg C) \\ &+ p(A) \cdot p(\neg C \mid A) \cdot p(B \mid A) \cdot p(\neg D \mid B, \neg C) \\ &+ p(A) \cdot p(C \mid \neg A) \cdot p(B \mid A) \cdot p(D \mid B, \neg C) \\ &+ p(\neg A) \cdot p(C \mid \neg A) \cdot p(B \mid \neg A) \cdot p(D \mid B,C) \\ &+ p(\neg A) \cdot p(\neg C \mid \neg A) \cdot p(B \mid \neg A) \cdot p(D \mid B, \neg C) \\ &+ p(\neg A) \cdot p(\neg C \mid \neg A) \cdot p(B \mid \neg A) \cdot p(\neg D \mid B, \neg C) \\ &+ p(\neg A) \cdot p(\neg C \mid \neg A) \cdot p(B \mid \neg A) \cdot p(\neg D \mid B, \neg C) \\ &= 0.7 \cdot 0.7 \cdot 0.2 \cdot 0.9 \\ &+ 0.7 \cdot 0.3 \cdot 0.8 \cdot 0.1 \\ &+ 0.7 \cdot 0.3 \cdot 0.2 \cdot 0.5 \\ &+ 0.7 \cdot 0.3 \cdot 0.2 \cdot 0.5 \\ &+ 0.3 \cdot 0.6 \cdot 0.4 \cdot 0.1 \\ &+ 0.3 \cdot 0.4 \cdot 0.4 \cdot 0.5 \\ &+ 0.3 \cdot 0.4 \cdot 0.4 \cdot 0.5 \\ &= 0.27 \end{split}$$

and p(D):

$$\begin{split} p(D) &= \sum_{A,C,B} p(A,B,C,D=1) \\ &= p(A) \cdot p(C \mid A) \cdot p(B \mid A) \cdot p(D \mid B,C) \\ &+ p(A) \cdot p(C \mid A) \cdot p(\neg B \mid A) \cdot p(D \mid \neg B,C) \\ &+ p(A) \cdot p(\neg C \mid A) \cdot p(B \mid A) \cdot p(D \mid B,\neg C) \\ &+ p(A) \cdot p(\neg C \mid A) \cdot p(\neg B \mid A) \cdot p(D \mid \neg B,\neg C) \\ &+ p(A) \cdot p(C \mid \neg A) \cdot p(B \mid \neg A) \cdot p(D \mid B,C) \\ &+ p(\neg A) \cdot p(C \mid \neg A) \cdot p(B \mid \neg A) \cdot p(D \mid B,C) \\ &+ p(\neg A) \cdot p(\neg C \mid \neg A) \cdot p(B \mid \neg A) \cdot p(D \mid B,\neg C) \\ &+ p(\neg A) \cdot p(\neg C \mid \neg A) \cdot p(B \mid \neg A) \cdot p(D \mid B,\neg C) \\ &+ p(\neg A) \cdot p(\neg C \mid \neg A) \cdot p(\neg B \mid \neg A) \cdot p(D \mid \neg B,\neg C) \\ &= 0.7 \cdot 0.7 \cdot 0.2 \cdot 0.9 \\ &+ 0.7 \cdot 0.3 \cdot 0.8 \cdot 0.3 \\ &+ 0.7 \cdot 0.3 \cdot 0.8 \cdot 0.3 \\ &+ 0.3 \cdot 0.6 \cdot 0.4 \cdot 0.9 \\ &+ 0.3 \cdot 0.4 \cdot 0.4 \cdot 0.5 \\ &+ 0.3 \cdot 0.4 \cdot 0.6 \cdot 0.3 \\ &= 0.36 \end{split}$$

# Exercise 2

(a)

$$p(A, B, C, D, E, F) = p(A) \cdot p(B \mid A) \cdot p(C \mid B) \cdot p(D \mid B) \cdot p(E \mid C, D) \cdot p(F \mid E)$$

(b) 
$$p(A, B, C, D, E) = p(A) \cdot p(B \mid A) \cdot p(C \mid B) \cdot p(D \mid B) \cdot p(E \mid C, D)$$

(c)

**Task:** Find a minimal set S that d-seperates A and F and prove that this is the case.

*Proof.* We propose  $S = \{B\}$  to prove that the criteria of S holds we have actually to prove two statements:

- 1. S d-seperates A and F
- 2. there is no set with fewer elements that also d-seperates A and F

We will start with the first statement. Between A and F exist two paths:

$$p_1 = A \to B \to C \to E \to F$$
  
 $p_2 = A \to B \to D \to E \to F$ 

We have  $p_1$  is blocked by S because with  $i_k = B$  we have :  $i_k \in S$  and  $A \to B \to C \leftrightarrow i_{k-1} \to i_k \to i_{k+1}$  and  $p_2$  is blocked by S because with  $i_k = B$  we have :  $i_k \in S$  and  $A \to B \to D \leftrightarrow i_{k-1} \to i_k \to i_{k+1}$  It is left to show that there is no set with fewer elements that also d-seperates A and F. The only set that has fewer elements is the empty set but the empty set is neither blocking  $p_1$  nor  $p_2$  according to the definition.

(d)

*Proof.* We will proof that  $C \perp\!\!\!\perp_G D \mid B$  i.e B d-seperates C and D holds. This is the case if every path is blocked by  $S = \{B\}$ . There are two paths from C to D:

$$p_1 = C \leftarrow B \rightarrow D$$
$$p_2 = C \rightarrow E \leftarrow D$$

 $p_1$  is blocked by S because with  $i_k = B$  we have  $i_k \in S$  and  $C \leftarrow B \rightarrow D \leftrightarrow i_{k-1} \leftrightarrow \leftarrow i_k \rightarrow i_{k+1}$ . Also  $p_2$  is blocked by S because with  $i_k = E$  we have :  $i_k \notin S$  and  $C \rightarrow E \leftarrow D \leftrightarrow i_{k-1} \rightarrow i_k \leftarrow i_{k+1}$  So every path between C and D is blocked by S and therefore  $C \perp_{G} D \mid B$  holds.

#### Exercise 3

(a)

We will proof that

$$A \to B \to C \implies A \perp\!\!\!\perp C \mid B$$

Proof.

# (b)

We will proof that

$$A \leftarrow B \leftarrow C \implies A \perp \!\!\! \perp \!\!\! \perp C \mid B$$

and we know from a that  $p(C) \cdot p(B \mid C) \Leftrightarrow p(B) \cdot p(C \mid B)$ .

Proof.

$$\begin{split} p(A,B,C) &= p(A\mid B) \cdot p(B\mid C) \cdot p(C) \\ \Leftrightarrow p(A,B,C) &= p(A\mid B) \cdot p(C\mid B) \cdot p(B) \\ \Leftrightarrow \frac{p(A,B,C)}{p(B)} &= p(A\mid B) \cdot p(C\mid B) \\ \Leftrightarrow p(A,C\mid B) &= p(A\mid B) \cdot p(C\mid B) \\ \Longrightarrow A \bot C \mid B \end{split}$$

### (c)

We will proof that

$$A \leftarrow B \rightarrow C \implies A \perp \!\!\! \perp \!\!\! \perp C \mid B$$

Proof.

$$\begin{split} p(A,B,C) &= p(A\mid B) \cdot p(B) \cdot p(C\mid B) \\ \Leftrightarrow \frac{p(A,B,C)}{p(B)} &= p(A\mid B) \cdot p(C\mid B) \\ \Leftrightarrow p(A,C\mid B) &= p(A\mid B) \cdot p(C\mid B) \\ \Longrightarrow A \! \perp \! \perp \! C \mid B \end{split}$$

### (d)

We will proof that

$$A \to B \leftarrow C \implies A \perp\!\!\!\perp C$$

Proof.

$$\begin{split} p(A,B,C) &= p(A) \cdot p(B \mid A,C) \cdot p(C) \\ \Leftrightarrow \frac{p(A,B,C)}{p(B \mid A,C)} &= p(A) \cdot P(C) \\ \Leftrightarrow \frac{p(A,B,C)}{\frac{p(A,B,C))}{p(A,C)}} &= p(A) \cdot p(C) \\ \Leftrightarrow p(A,C) &= p(A) \cdot p(C) \\ \Leftrightarrow A \bot C \end{split}$$

# Exercise 4

(a)

We will proof that  $E(a \cdot X + Y) = a \cdot E(X) + E(Y)$  with the sum rule and the factor rule.

Proof.

$$E(a \cdot X + Y) = \int a \cdot X + Y dx$$
$$= \int a \cdot X dx + \int Y dx$$
$$= a \cdot \int X dx + \int Y dx$$
$$= a \cdot E(X) + E(Y)$$

(b)

We will proof that  $Var(a \cdot X) = a^2 \cdot Var(X)$  with the result of a.

Proof.

$$Var(a \cdot X) = E(a \cdot X - \mu)^{2}$$

$$= E(a^{2} \cdot (X - \mu)^{2})$$

$$= a^{2} \cdot E(X - \mu)^{2}$$

$$= a^{2} \cdot Var(X)$$

# Exercise 5

In this Exercise we attempt to show that the mean  $\mu$  is also the mode of the Gaussian normal distribution:

$$N(x \mid \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{\frac{-(x-u)^2}{2\sigma^2}}$$

*Proof.* The mode of N is the point where N obtains it's maximum value. We will therefore search for the point where N is maximal. This is the case only if  $\frac{dN}{dx}=0$ 

$$\frac{dN}{dx} = \frac{-2(x-u)}{2\sigma^2 \sqrt{2\pi\sigma^2}} e^{\frac{-(x-u)^2}{2\sigma^2}}$$

Since  $e^x>0$  for all x this function becomes only 0 if x=u. We can also deduct that for  $x\in (-\infty\mu]: \frac{-2(x-u)}{2\sigma^2\sqrt{2\pi\sigma^2}}\geq 0$  and for  $x\in [\mu,\infty): \frac{-2(x-u)}{2\sigma^2\sqrt{2\pi\sigma^2}}\leq 0$ , since  $e^x>0$  this determines the behaviour of the drivative and there for  $(\mu,N(\mu\mid\mu,\sigma))$  is the maximum.

# Exercise 6

Our Implementation is in the Jupyter Notebook "Exercise 6"