

# Copenhagen interview questions

January 13, 2022

**Exercise 1** - (Answer:  $\beta = (\frac{1}{3}, \frac{4}{3})$ )

a) Consider  $\mathbf{X} = (X_1, \dots, X_d)^T \in \mathbb{R}^d$  vector of random variables and  $Y \in \mathbb{R}$  (the response) a scalar random variable. Assume that all these random variables have mean zero and finite variance.

Let the covariance matrices be

$$\text{Cov}(\mathbf{X}, Y) = \begin{bmatrix} \text{Cov}(X_1, Y) \\ \vdots \\ \text{Cov}(X_d, Y) \end{bmatrix} \in \mathbb{R}^d \quad (1)$$

$$\text{Cov}(\mathbf{X}) = \begin{bmatrix} \text{Var}(X_1) & \dots & \text{Cov}(X_1, X_d) \\ \vdots & & \vdots \\ \text{Cov}(X_d, X_1) & \dots & \text{Var}(X_d) \end{bmatrix} \in \mathbb{R}^{d \times d} . \quad (2)$$

Given the zero mean assumption, covariance can be rewritten as follow:

$$\text{Cov}(X_i, X_j) = \mathbb{E}(X_i X_j), \quad \forall i, j \in [1, d] \quad (3)$$

$$\text{Cov}(X_i, Y) = \mathbb{E}(X_i Y), \quad \forall i \in [1, d] . \quad (4)$$

Finally, let us define the joint probability distribution as  $f(Y, \mathbf{X})$  and assume variables changing on a continuous range.

In order to find  $\beta$  minimizing  $\mathbb{E}([Y - \beta^t \mathbf{X}]^2)$ , we derive the expectation over  $\beta$  and set the result equals to zero (the critical point is a minimum since we are dealing with a parabola).

$$\frac{\partial \mathbb{E}([Y - \beta^T \mathbf{X}]^2)}{\partial \beta} = -2 \int_{\mathbf{x}} \int_y (y - \beta^T \mathbf{x})^T (y - \beta^T \mathbf{x}) f(\mathbf{x}, y) dy d\mathbf{x} = 0$$

$$\int_{\mathbf{x}} \int_y (y - \beta^T \mathbf{x}) \mathbf{x} f(\mathbf{x}, y) dy d\mathbf{x} = 0$$

$$\int_{\mathbf{x}} \int_y y \mathbf{x} f(\mathbf{x}, y) dy d\mathbf{x} = \beta^T \int_{\mathbf{x}} \mathbf{x}^T \mathbf{x} f(\mathbf{x}) d\mathbf{x}$$

$$\mathbb{E}(Y \mathbf{X}) = \beta^T \mathbb{E}(\mathbf{X}^T \mathbf{X})$$

$$\text{Cov}(Y, \mathbf{X}) = \beta^T \text{Cov}(\mathbf{X})$$

$$\beta^T = \text{Cov}(\mathbf{X})^{-1} \text{Cov}(Y, \mathbf{X}) , \quad (5)$$

where we supposed that  $\text{Cov}(\mathbf{X})$  is an invertible matrix, and we used Equations (4) and (3) to get covariance from expectations.

**b)** We use the formula found in part a) in order to solve this exercise. From the provided definition of  $X, Y, Z$  random variables, we have the following distributions:

$$X \sim N(0, 1) ,$$

$$Y \sim N(0, 2) ,$$

$$Z \sim N(0, 4) .$$

Given  $\mathbf{X} = (X, Z)$  the covariance matrices are:

$$\text{Cov}(\mathbf{X}) = \begin{pmatrix} \text{Var}(X) & \text{Cov}(X, Z) \\ \text{Cov}(X, Z) & \text{Var}(Z) \end{pmatrix} , \quad (6)$$

$$\text{Cov}(\mathbf{X}, Y) = \begin{pmatrix} \text{Cov}(X, Y) \\ \text{Cov}(Z, Y) \end{pmatrix} . \quad (7)$$

Let's do the math and find the required covariances using the same intuitions of Equations (3), (4) and using our knowledge on the variances.

$$\text{Cov}(X, Z) = \mathbb{E}(XZ) = -1 ,$$

$$\text{Cov}(X, Y) = \mathbb{E}(XY) = -1 ,$$

$$\text{Cov}(Z, Y) = \mathbb{E}(ZY) = 5 ,$$

which must be substituted in (6) and (7) leading to

$$\text{Cov}(\mathbf{X}) = \begin{pmatrix} 1 & -1 \\ -1 & 4 \end{pmatrix} , \quad (8)$$

$$\text{Cov}(\mathbf{X}, Y) = \begin{pmatrix} -1 \\ 5 \end{pmatrix} . \quad (9)$$

Using Gauss-Jordan method we start from  $(\text{Cov}(\mathbf{X}) \mid \mathbf{I})$  in order to find the inverse matrix

$$\text{Cov}(\mathbf{X})^{-1} = \begin{pmatrix} \frac{4}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} \end{pmatrix} \quad (10)$$

Finally we get the value of  $\beta$  minimizing the expectation of the squared error by solving Equation (5):

$$\begin{aligned} \beta &= \text{Cov}(\mathbf{X})^{-1} \text{Cov}(Y, \mathbf{X}) = \\ &= \begin{pmatrix} \frac{4}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} \end{pmatrix} \begin{pmatrix} -1 \\ 5 \end{pmatrix} = \begin{pmatrix} \frac{1}{3} \\ \frac{4}{3} \end{pmatrix} \end{aligned}$$

## Exercise 2 - (Answer: 9)

Given the set of vertices  $V = \{A, B, C, D\}$ , the DAGs satisfying the problem's requirements are constrained as follow:

- No cycles in the graph
- The only d-separated nodes are  $A$  and  $C$ .

Since we are not conditioning on any variable, then  $A \perp\!\!\!\perp C \mid \emptyset$  in the following cases:

- Node  $B$  is a collider relative to the path from  $A$  to  $C$  and every path is blocked by  $B$
- Node  $D$  is a collider relative to the path from  $A$  to  $C$  and every path is blocked by  $D$

- Both nodes  $B$  and  $D$  are colliders relative to the path from  $A$  to  $C$ .

Note the symmetry of the problem with respect to  $B$  and  $D$ : once we find the DAGs where  $A \perp\!\!\!\perp C \mid \emptyset$  because every path is blocked by  $B$ , by replacing  $B$  with  $D$  we will simply double the number of DAGs satisfying the condition. There are also 3 additional graphs where both  $B$  and  $D$  are colliders and must be considered in the count.

Following the above reasoning, we can find 9 DAGs where  $A \perp\!\!\!\perp C \mid \emptyset$  and no other group of vertices is d-separated.

### Exercise 3

#### Example 1

$$A = \begin{pmatrix} 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 \end{pmatrix}$$

- 1: 2, 3, 4  
 2: 4  
 3: 4  
 4:  
 5: 3, 4

#### Example 2

$$A = \begin{pmatrix} 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

- 1: 2, 3, 4, 5  
 2:  
 3: 5  
 4: 5  
 5: