

## Probability theory and Linear Algebra

**Question 1:** Some probability basics

(a) How many different  $\sigma$ -algebras is it possible to define on the set  $\Omega = \{A, B, C\}$ ?

Specifically write down every possibility.

(b) Prove that the pdf of the Binomial distribution indeed qualifies as a probability function.

*Hint:* You may use the Binomial theorem, which states that

$$(a + b)^n = \sum_{i=0}^n \binom{n}{i} a^i b^{n-i}. \quad (1)$$

(c) Consider the following function

$$F : \mathbb{R} \rightarrow \mathbb{R}, \quad x \mapsto \begin{cases} 1, & x \geq d \\ \frac{1}{3}x^3 + \frac{1}{6}x + \frac{1}{2}, & -d \leq x \leq d \\ 0, & x \leq -d \end{cases}$$

for some  $d \in \mathbb{R}_{>0}$ .

(i) Determine a value for  $d$  so that  $F$  is a CDF.

(ii) Write down the corresponding probability (density) function and calculate  $P(X \in A)$  for a random variable  $X$  with CDF  $F$  and

$$A = \{\{x \in \mathbb{R} \mid x > 0\} \cup \{x \in \mathbb{R} \mid x \leq -d\}\}^C.$$

(d) Consider a test to detect a disease that 0.1% of the population have. The test is 99% effective in detecting an infected person. However, the test gives a false positive result in 0.5% of cases. If a person tests positive for the disease what is the probability that they actually have it?

### Solution:

(a) There are 5 possibilities for a  $\sigma$ -algebra on  $\Omega$ :

(i)  $\{\emptyset, \Omega\}$

(ii)  $\{\emptyset, \{A\}, \{B, C\}, \Omega\}$

(iii)  $\{\emptyset, \{B\}, \{A, C\}, \Omega\}$

(iv)  $\{\emptyset, \{C\}, \{A, B\}, \Omega\}$

(v)  $\{\emptyset, \{A\}, \{B\}, \{C\}, \{B, C\}, \{A, C\}, \{A, B\}, \Omega\} = \mathcal{P}(\Omega)$

(b) Given that the probability function of the Binomial distribution is  $\binom{n}{x} p^x (1-p)^{n-x}$  with

support  $\{0, 1, \dots, n\} \forall n \in \mathbb{N}$ , this immediately follows from:

$$1 = (p + 1 - p)^n = \sum_{x=0}^n \binom{n}{x} p^x (1-p)^{n-x} \quad \forall n \in \mathbb{N}.$$

(c) (i) We require that

$$\int_{-\infty}^{\infty} F'(x) dx = \int_{-d}^d F'(x) dx \stackrel{!}{=} 1,$$

and, therefore,

$$\begin{aligned} \int_{-\infty}^{\infty} F'(x) dx &= F(d) - F(-d) \\ &= \frac{1}{3}d^3 + \frac{1}{6}d + \frac{1}{2} - \left[ \frac{1}{3}(-d)^3 + \frac{1}{6}(-d) + \frac{1}{2} \right] \\ &= \frac{2}{3}d^3 + \frac{1}{3}d \stackrel{!}{=} 1. \end{aligned}$$

It follows that  $d = 1$ .

(ii) The corresponding density is given by

$$f(x) = \begin{cases} 0, & x > 1 \\ F'(x) = x^2 + \frac{1}{6}, & -1 \leq x \leq 1 \\ 0, & x \leq -1. \end{cases}$$

Meanwhile, we have that

$$\begin{aligned} P(X \in \{\{x \in \mathbb{R} \mid x > 0\} \cup \{x \in \mathbb{R} \mid x \leq -d\}\}^C) &= P(X \in \{x \in \mathbb{R} \mid x > 0\}^C \cap \{x \in \mathbb{R} \mid x \leq -d\}^C) \\ &= P(X \in \{x \in \mathbb{R} \mid x \leq 0\} \cap \{x \in \mathbb{R} \mid x > -d\}) \\ &= P(X \in ]-d, 0]) = P(X \leq 0) - P(X \leq -d) \\ &= F(0) - F(-d) = \frac{1}{2}. \end{aligned}$$

(d) The first step in solving this problem is to recognize what we are trying to calculate and what quantities we have been given in the question. We can introduce a Bernoulli random variable and say that  $D = 1$  when the person has the disease. We can then introduce a second Bernoulli random variable  $T$  and say that  $T = 1$  when a person gets a positive test result. We then note that these two random variables are **not** independent. With these symbols in place we can now state clearly what it we are trying to calculate. We are trying to calculate the conditional probability  $P(D = 1 \mid T = 1)$ . In addition, the question tells us that:

$$P(D = 1) = 0.001 \quad P(T = 1 \mid D = 1) = 0.990 \quad P(T = 1 \mid D = 0) = 0.005$$

From these quantities we can calculate the probability of getting a positive test result,  $P(T = 1)$  using

$$\begin{aligned} P(T = 1) &= P(T = 1 \mid D = 1)P(D = 1) + P(T = 1 \mid D = 0)P(D = 0) \\ &= P(T = 1 \mid D = 1)P(D = 1) + P(T = 1 \mid D = 0)[1 - P(D = 1)] \\ &= 0.99 \times 0.1 + 0.005 \times (1 - 0.001) = 0.005985 \end{aligned}$$

We can now insert this result into Bayes theorem to get the desired conditional probability.

$$P(D = 1|T = 1) = \frac{P(T = 1|D = 1)P(D = 1)}{P(T = 1)} = \frac{0.99 \times 0.001}{0.005985} \approx 0.165.$$

**Question 2:** Matrix rank and linear independence

(a) Calculate the rank of the following matrix:

$$\begin{bmatrix} 4 & -6 & 0 \\ -6 & 0 & 1 \\ 0 & 9 & -1 \\ 0 & 1 & 4 \end{bmatrix}$$

(b) Are the following vectors linearly independent?

$$v_1 = \begin{pmatrix} 3 \\ 0 \\ 1 \\ 2 \end{pmatrix}; v_2 = \begin{pmatrix} 6 \\ 1 \\ 0 \\ 0 \end{pmatrix}; v_3 = \begin{pmatrix} 12 \\ 1 \\ 2 \\ 4 \end{pmatrix}; v_4 = \begin{pmatrix} 6 \\ 0 \\ 2 \\ 4 \end{pmatrix}; v_5 = \begin{pmatrix} 9 \\ 0 \\ 1 \\ 2 \end{pmatrix}$$

If they are not, find the largest number of linearly independent vectors among them.

(c) Prove that if a matrix  $\mathbf{A} \in \mathbb{R}^{n \times m}$  is not square, then either the row vectors or the column vectors are linearly dependent.

**Solution:**

(a)

$$\begin{aligned} \begin{bmatrix} 4 & -6 & 0 \\ -6 & 0 & 1 \\ 0 & 9 & -1 \\ 0 & 1 & 4 \end{bmatrix} &\xrightarrow{R1 \cdot (-1) - R2 - R3} \begin{bmatrix} 2 & -3 & 0 \\ -6 & 0 & 1 \\ 0 & 9 & -1 \\ 0 & 1 & 4 \end{bmatrix} \\ &\xrightarrow{R2+3R1} \begin{bmatrix} 2 & -3 & 0 \\ 0 & -9 & 1 \\ 0 & 9 & -1 \\ 0 & 1 & 4 \end{bmatrix} \\ &\xrightarrow{R4+R3; R5+\frac{1}{9}R2} \begin{bmatrix} 2 & -3 & 0 \\ 0 & -9 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 37/9 \end{bmatrix} \\ &\xrightarrow{R4+R5; R5-R5} \begin{bmatrix} 2 & -3 & 0 \\ 0 & -9 & 1 \\ 0 & 0 & 37/9 \\ 0 & 0 & 0 \end{bmatrix} \end{aligned}$$

Therefore, the rank of the original matrix is 3 .

(b) This question is equivalent to asking for the rank of the matrix

$$\begin{bmatrix} 3 & 0 & 1 & 2 \\ 6 & 1 & 0 & 0 \\ 12 & 1 & 2 & 4 \\ 6 & 0 & 2 & 4 \\ 9 & 0 & 1 & 2 \end{bmatrix} \implies \begin{bmatrix} 3 & 0 & 1 & 2 \\ 0 & 1 & -2 & -4 \\ 0 & 1 & -2 & -4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & -2 & -4 \end{bmatrix} \implies \begin{bmatrix} 3 & 0 & 1 & 2 \\ 0 & 1 & -2 & -4 \\ 0 & 0 & -2 & -4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

The rank of the matrix is 3. It follows that the maximum number of linearly independent vectors is also 3. They are the ones that correspond to the non-zero rows of the final matrix:

$$(3, 0, 1, 2)^\top; (6, 1, 0, 0)^\top; (9, 0, 1, 2)^\top.$$

(c) The maximum number of linearly independent row vectors is the rank of  $\mathbf{A}$ , while the maximum number of linearly independent column vectors is the rank of  $\mathbf{A}^\top$ . If  $n < m$ , then  $\text{rank}(\mathbf{A}^\top) = \text{rank}(\mathbf{A}) \leq n < m$ . Therefore, the column vectors are linearly dependent. Similarly, if  $m < n$ , then the row vectors are linearly dependent.

### Question 3: Matrixdecomposition

Consider the following matrices:

$$\mathbf{A} = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}, \quad \mathbf{B} = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}, \quad \mathbf{C} = \begin{pmatrix} 1 & 2 \\ 1 & 2 \end{pmatrix} \quad \text{und} \quad \mathbf{D} = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 5 & 7 \\ 3 & 7 & 26 \end{pmatrix}.$$

You may assume it to be known that the matrix  $\mathbf{D}$  is positive definite.

- Determine the definitiveness of  $\mathbf{A}$ ,  $\mathbf{B}$  and  $\mathbf{C}$ .
- Decompose the matrix  $\mathbf{A}$  using Eigendecomposition.
- Determine the entries of  $\mathbf{L}$  in the Cholesky-decomposition  $\mathbf{D} = \mathbf{L}\mathbf{L}^\top$  of  $\mathbf{D}$ .

### Solution:

a) **Definiteness:**

$$\mathbf{A} = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$$

characteristic polynomial:

$$\det(\mathbf{A} - \lambda\mathbf{I}) = \det \begin{pmatrix} 2 - \lambda & 1 \\ 1 & 2 - \lambda \end{pmatrix} = (2 - \lambda)^2 - 1^2 \stackrel{!}{=} 0$$

$$(2 - \lambda)^2 = 1$$

$$2 - \lambda = \pm 1$$

$$\Rightarrow \lambda_1 = 1 \quad \lambda_2 = 3$$

$\Rightarrow \mathbf{A}$  is positive definite, since all eigenvalues are greater than zero!

$$\mathbf{B} = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}$$

characteristic polynomial:

$$\begin{aligned} \det(\mathbf{B} - \lambda\mathbf{I}) &= \det \begin{pmatrix} 1-\lambda & 2 \\ 2 & 1-\lambda \end{pmatrix} = (1-\lambda)^2 - 2^2 \stackrel{!}{=} 0 \\ (1-\lambda)^2 &= 4 \\ 1-\lambda &= \pm 2 \\ \Rightarrow \lambda_1 = 3 &\quad \lambda_2 = -1 \end{aligned}$$

$\Rightarrow \mathbf{B}$  is indefinite, since one eigenvalue is smaller and one greater than zero!

$$\mathbf{C} = \begin{pmatrix} 1 & 2 \\ 1 & 2 \end{pmatrix}$$

$\mathbf{C}$  is not symmetric. Therefore, definiteness cannot be directly determined through eigenvalues. However, we can use the eigenvalues of the corresponding symmetric matrix! Simply put: we keep the diagonal, but take the pairwise average over the elements of the off-diagonal.

$$\mathbf{C}_S = \frac{1}{2} (\mathbf{C} + \mathbf{C}^\top).$$

$$\mathbf{C}_S = \frac{1}{2} \left[ \begin{pmatrix} 1 & 2 \\ 1 & 2 \end{pmatrix} + \begin{pmatrix} 1 & 1 \\ 2 & 2 \end{pmatrix} \right] = \begin{pmatrix} 1 & 1.5 \\ 1.5 & 2 \end{pmatrix}$$

characteristic polynomial:

$$\begin{aligned} \det(\mathbf{C}_S - \lambda\mathbf{I}) &= \det \begin{pmatrix} 1-\lambda & 1.5 \\ 1.5 & 2-\lambda \end{pmatrix} \stackrel{!}{=} 0 \\ &\Rightarrow (1-\lambda)(2-\lambda) - 1.5^2 \\ &= \lambda^2 - 3\lambda - 0.25 \stackrel{!}{=} 0 \\ \Rightarrow \lambda_{1/2} &= \frac{3 \pm \sqrt{10}}{2} \\ \lambda_1 &\approx -0.08 \quad \lambda_2 \approx 3.08 \end{aligned}$$

$\Rightarrow \mathbf{C}$  is indefinite, since one eigenvalue is smaller and one greater than zero!

b) **Eigendecomposition:**

$$\mathbf{A} = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \quad \text{with eigenvalue } \lambda_1 = 1, \lambda_2 = 3$$

Calculating the eigenvectors:

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix} \stackrel{!}{=} (\mathbf{A} - \lambda_1 \mathbf{I}) \cdot \mathbf{x} = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 + x_2 \\ x_1 + x_2 \end{pmatrix}$$

$$\Rightarrow x_2 = -x_1 \Rightarrow \text{normed eigenvector: } \sqrt{1/2} \cdot \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix} \stackrel{!}{=} (\mathbf{A} - \lambda_2 \mathbf{I}) \cdot \mathbf{x} = \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} -x_1 + x_2 \\ x_1 - x_2 \end{pmatrix}$$

$$\Rightarrow x_2 = x_1 \Rightarrow \text{normed eigenvector: } \sqrt{1/2} \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\Rightarrow \mathbf{A} = \underbrace{\sqrt{1/2} \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix}}_{\mathbf{P}} \cdot \underbrace{\begin{pmatrix} 1 & 0 \\ 0 & 3 \end{pmatrix}}_{\text{diag}(\lambda_1, \lambda_2)} \cdot \underbrace{\sqrt{1/2} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}}_{\mathbf{P}^T}$$

c) Cholesky-decomposition:

$$\mathbf{D} = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 5 & 7 \\ 3 & 7 & 26 \end{pmatrix} = \begin{pmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & l_{33} \end{pmatrix} \begin{pmatrix} l_{11} & l_{21} & l_{31} \\ 0 & l_{22} & l_{32} \\ 0 & 0 & l_{33} \end{pmatrix} = \mathbf{L}\mathbf{L}^T$$

Calculating the Cholesky decomposition:

$$l_{ii} = (d_{ii} - \sum_{k=1}^{i-1} l_{ik}^2)^{\frac{1}{2}}$$

$$l_{ji} = \frac{1}{l_{ii}} (d_{ji} - \sum_{k=1}^{i-1} l_{jk} l_{ik})$$

Specifically:

$$l_{11} = (d_{11} - \sum_{k=1}^{1-1} l_{1k}^2)^{\frac{1}{2}} = (1 - 0)^{\frac{1}{2}} = 1$$

$$l_{21} = \frac{1}{l_{11}} (d_{21} - \sum_{k=1}^{1-1} l_{2k} l_{1k}) = \frac{d_{21}}{l_{11}} = \frac{2}{1} = 2$$

$$l_{31} = \frac{1}{l_{11}} (d_{31} - \sum_{k=1}^{1-1} l_{3k} l_{1k}) = \frac{d_{31}}{l_{11}} = \frac{3}{1} = 3$$

$$l_{22} = (d_{22} - \sum_{k=1}^{2-1} l_{2k}^2)^{\frac{1}{2}} = (5 - 2^2)^{\frac{1}{2}} = 1$$

$$l_{32} = \frac{1}{l_{22}} (d_{32} - \sum_{k=1}^{2-1} l_{3k} l_{2k}) = \frac{1}{1} (7 - 3 \cdot 2) = 1$$

$$l_{33} = (d_{33} - \sum_{k=1}^{3-1} l_{3k}^2)^{\frac{1}{2}} = (d_{33} - (l_{31}^2 + l_{32}^2)^{\frac{1}{2}}) = (26 - 3^2 - 1^2)^{\frac{1}{2}} = 16^{\frac{1}{2}} = 4$$

$$\Rightarrow \mathbf{L} = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 1 & 4 \end{pmatrix} \quad \text{und} \quad \mathbf{D} = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 5 & 7 \\ 3 & 7 & 26 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 1 & 4 \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 0 & 0 & 4 \end{pmatrix}.$$