

Final Exam (2h)

The exam is open book.

Exercise 1 [Random vectors]

Let (X, Y) have joint density

$$f(x, y) = \begin{cases} \frac{2x+4y}{3} & \text{if } (x, y) \in [0, 1]^2 \\ 0 & \text{otherwise.} \end{cases}$$

1. Find the marginal densities f_X and f_Y of X and Y .
2. Are X and Y independent? Justify briefly.
3. Prove that

$$\mathbb{E}[X|Y] = \frac{2 + 6Y}{3 + 12Y}.$$

Answers :

1. Let us compute, for a fixed $x \in [0, 1]$,

$$f_X(x) = \int_{y=0}^1 \frac{2x+4y}{3} dy = \frac{2x}{3} + \int_{y=0}^1 \frac{4}{3} y dy = \frac{2x}{3} + \frac{2}{3}.$$

Similarly, for a fixed $y \in [0, 1]$

$$f_Y(y) = \int_{x=0}^1 \frac{2x+4y}{3} dx = \frac{4y}{3} + \int_{x=0}^1 \frac{2}{3} x dx = \frac{4y}{3} + \frac{1}{3}.$$

2. If X, Y were independent, then one would have $f(x, y) = f_X(x)f_Y(y)$, this is not the case.
3. According to the course formula (p.37),

$$\begin{aligned} \mathbb{E}[X|Y] &= \frac{\int_x x f(x, Y) dx}{f_Y(Y)} \\ &= \frac{\int_x x(2x + 4Y)/3 dx}{4Y/3 + 1/3} \\ &= \frac{\int_x x(2x + 4Y) dx}{4Y + 1} \\ &= \frac{2 \int_{x=0}^1 x^2 dx + 4Y \int_{x=0}^1 x dx}{4Y + 1} \\ &= \frac{2/3 + 4Y/2}{4Y + 1} = \frac{2 + 6Y}{3 + 12Y}. \end{aligned}$$

Exercise 2 [Random variables]

Let X have density $f_X(x) = \begin{cases} \frac{3}{x^4} & \text{if } x \geq 1, \\ 0 & \text{otherwise.} \end{cases}$

1. Compute $\mathbb{E}[X^2]$. Does X belong to L^2 ?
2. Compute the cumulative distribution function F_X of X .
3. Prove that $1 - \frac{1}{X^3}$ have the uniform distribution over $[0, 1]$.
(Hint : you can use the CDF, or the "change of variable" method.)

Answers :

1.

$$\mathbb{E}[X^2] = \int_1^{+\infty} x^2 \frac{3}{x^4} dx = -3/x \Big|_1^\infty = 3 < +\infty,$$

therefore $X \in L^2$.

2. Let $t \geq 1$,

$$F_X(t) = \mathbb{P}(X \leq t) = \int_1^t \frac{3}{x^4} dx = -1/x^3 \Big|_1^t = 1 - \frac{1}{t^3}.$$

3. We first observe that $X \in [1, +\infty)$, so $1/X^3 \in [0, 1]$, and $1 - 1/X^3 \in [0, 1]$ also.

1st method : CDF. Let us compute its CDF : for any $0 \leq t \leq 1$

$$\begin{aligned} \mathbb{P}(1 - 1/X^3 \leq t) &= \mathbb{P}(-1/X^3 \leq t - 1) \\ &= \mathbb{P}(X^3 \leq 1/(1 - t)) \\ &= \mathbb{P}\left(X \leq (1/(1 - t))^{1/3}\right) \\ &= F_X\left((1/(1 - t))^{1/3}\right) \\ &= 1 - \frac{1}{((1/(1 - t))^{1/3})^3} = 1 - \frac{1}{1/(1 - t)} = 1 - (1 - t) = t. \end{aligned}$$

Now, $F(t) = t$ is the CDF of the uniform distribution in $[0, 1]$.

2d method : "change of variable". Let ϕ be any continuous and bounded function, and let us compute

$$\mathbb{E}[\phi(1 - 1/X^3)] = \int_1^{+\infty} \phi(1 - 1/x^3) 3/x^4 dx.$$

We put $u = 1 - 1/x^3$, then $x = \frac{1}{(1-u)^{1/3}}$, and $\frac{du}{dx} = \frac{3}{x^4}$. The boundaries become : $x = 1 \Leftrightarrow u = 0$ and $x = +\infty \Leftrightarrow u = 1$. Thus,

$$\mathbb{E}[\phi(1 - 1/X^3)] = \int_{u=0}^1 \phi(u) du = \int_u \phi(u) \mathbf{1}_{[0,1]}(u) du,$$

which proves that $1 - 1/X^3$ is uniform (Theorem 2.7).

Exercise 3 [Gaussian Vectors]

Let $\begin{pmatrix} X \\ Y \end{pmatrix}$ be a gaussian vector with mean $\mu = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ and with covariance matrix

$$C = \begin{pmatrix} 1 & -1 \\ -1 & 3 \end{pmatrix}.$$

1. Briefly explain why $\begin{pmatrix} aY \\ X - aY \end{pmatrix}$ is a gaussian vector.

2. Let a be a fixed constant. Compute $\text{Cov}(Y, X - aY)$, and deduce the value of a such that Y and $X - aY$ are independent.

3. Compute $\mathbb{E}[X|Y]$.

(Hint : Write $\mathbb{E}[X|Y] = \mathbb{E}[aY + (X - aY)|Y]$, where a is the answer to the previous question.)

Correction :

1. The vector $\begin{pmatrix} aY \\ X - aY \end{pmatrix}$ is the image of $\begin{pmatrix} X \\ Y \end{pmatrix}$ by a linear transformation :

$$\begin{pmatrix} aY \\ X - aY \end{pmatrix} = \begin{pmatrix} 0 & a \\ 1 & -a \end{pmatrix} \begin{pmatrix} X \\ Y \end{pmatrix}.$$

Thus, it is a gaussian vector (see Proposition 4.5).

2.

$$\text{Cov}(Y, X - aY) = \text{Cov}(Y, X) - a\text{Cov}(Y, Y) = -1 - 3a.$$

Therefore we have $\text{Cov}(Y, X - aY) = 0$ for $a = -1/3$. Now, Y and $X - aY = X + Y/3$ are the two components of a gaussian vector with a null covariance. Therefore they are independent (Proposition 4.4).

3. Since $X = -Y/3 + (X + Y/3)$, we have by linearity of the conditional expectation

$$\begin{aligned}\mathbb{E}[X|Y] &= \mathbb{E}[-Y/3|Y] + \mathbb{E}[X + Y/3|Y] \\ &= -Y/3 + \mathbb{E}[X + Y/3] \\ &\quad \text{('taking out what is known')} \quad \text{(independence)}\end{aligned}$$

Finally, $\mathbb{E}[X + Y/3] = \mathbb{E}[X] + \mathbb{E}[Y/3] = 0 + 0$ since $\binom{aY}{X-aY}$ has mean $\binom{0}{0}$. We get

$$\mathbb{E}[X|Y] = -Y/3.$$

Exercise 4 [A kind of Law of Large Numbers]

Let $(X_k)_{k \geq 1}$ be a sequence of independent random variables such that for all $k \geq 1$,

$$X_k = \begin{cases} \sqrt{k} & \text{with probability } \frac{1}{\sqrt{k}} \\ 0 & \text{with probability } 1 - \frac{1}{\sqrt{k}}. \end{cases}$$

We set $S_n = X_1 + \dots + X_n$.

1. Compute, for every k , $\mathbb{E}[X_k]$, $\mathbb{E}[S_n]$, $\text{Var}[X_k]$, $\text{Var}[S_n]$.

The goal of the remainder of the exercise is to prove that

$$\frac{S_n}{n} \xrightarrow{\text{(prob.)}} 1.$$

Yet, the law of large numbers cannot be applied here, since X_k 's have not the same distribution.

2. Check that for every n we have $\text{Var}[S_n] \leq n\sqrt{n}$.

3. Using the Chebychev's inequality (page 15), prove that $\frac{S_n}{n}$ converges to 1 in probability.

Answers :

1. We have

$$\begin{aligned}\mathbb{E}[X_k] &= \sqrt{k} \times \frac{1}{\sqrt{k}} + 0 \times \left(1 - \frac{1}{\sqrt{k}}\right) = 1, \\ \mathbb{E}[S_n] &= \mathbb{E}[X_1] + \dots + \mathbb{E}[X_n] = n \times 1 = n.\end{aligned}$$

Besides,

$$\mathbb{E}[X_k^2] = (\sqrt{k})^2 \times \frac{1}{\sqrt{k}} = \sqrt{k}.$$

We get $\text{Var}[X_k] = \mathbb{E}[X_k^2] - \mathbb{E}[X_k]^2 = \sqrt{k} - 1$.

Since X_k 's are independent,

$$\begin{aligned}\text{Var}[S_n] &= \text{Var}[X_1] + \dots + \text{Var}[X_n] \\ &= \sqrt{1} - 1 + \sqrt{2} - 1 + \dots + \sqrt{n} - 1 = \sqrt{1} + \sqrt{2} + \dots + \sqrt{n} - n.\end{aligned}$$

2. We have

$$\begin{aligned}\text{Var}[S_n] &= \sqrt{1} + \sqrt{2} + \dots + \sqrt{n} - n \\ &\leq \sqrt{1} + \sqrt{2} + \dots + \sqrt{n} \\ &\leq \sqrt{n} + \sqrt{n} + \dots + \sqrt{n} \\ &= n\sqrt{n}.\end{aligned}$$

3. We need to prove that for every $\varepsilon > 0$,

$$\mathbb{P}\left(\left|\frac{S_n}{n} - 1\right| \geq \varepsilon\right) \xrightarrow{n \rightarrow +\infty} 0.$$

Using the Chebychev's inequality

$$\begin{aligned}\mathbb{P}\left(\left|\frac{S_n}{n} - 1\right| \geq \varepsilon\right) &= \mathbb{P}\left(\left|\frac{S_n}{n} - \mathbb{E}\left[\frac{S_n}{n}\right]\right| \geq \varepsilon\right) \\ &\leq \frac{\text{Var}(S_n/n)}{\varepsilon^2} \\ &= \frac{\text{Var}(S_n)/n^2}{\varepsilon^2} \quad (\text{see equation (\$) page 14}) \\ &\leq \frac{\sqrt{n}/n}{\varepsilon^2} \quad (\text{previous question}) \\ &\xrightarrow{n \rightarrow +\infty} 0.\end{aligned}$$