# PC 1 – Sets, Measures and Random Variables

## Set theory

#### Exercise 1

For  $n \geq 1$ , let

$$A_n = \left[ -\frac{1}{n}; 2 + \frac{1}{n} \right], \quad B_n = \left[ -\frac{5}{n}; n^2 \right].$$

1. Compute  $\bigcup_{n\geq 1} A_n, \bigcap_{n\geq 1} A_n$  and  $\limsup_n A_n$ , where  $\limsup_n A_n$  is defined as

 $\limsup A_n = \bigcap_{n \geq 1} \bigcup_{k \geq n} A_k = \left\{ x \text{ such that " } x \in A_n \text{ for infinitely many } n \text{ " } \right\}.$ 

- 2. Compute  $\bigcup_{n>1} B_n$ ,  $\bigcap_{n>1} B_n$  and  $\limsup_n B_n$ .
- 3. Evaluate the following set

$$\left\{ x \text{ such that } \sum_{n \ge 1} \mathbf{1}_{A_n}(x) = +\infty \right\}.$$

 $\textbf{Solution 1} \qquad \textit{1. Rappel de la définition}: \limsup_{n \to \infty} A_n = \cap_{k \geq 1} \cup_{n \geq k} A_n. \ \textit{Il s'agit de l'événement où}:$ 

$$\omega \in \limsup_{n \to \infty} A_n \iff il \ existe \ une \ infinit\'e \ de \ n \ tels \ que \ \omega \in A_n.$$

La suite  $(A_n)_{n\geq 1}$  étant monotone décroissante  $(A_n\supset A_{n+1} \text{ pour tout } n\geq 1)$ , on a pour tout  $k\geq 1$ ,  $\cup_{n\geq k}A_n=[-1/k,3+1/k]$ . D'une part, on voit que  $[0,3]\subset A_k\subset \cup_{n\geq k}A_n$  pour tout k. D'autre part, pour tout s<0 et pour tout t>3 il existe k tel que s<-1/k et t>3+1/k. Donc,  $\limsup_{n\to\infty}A_n=[0,3]$ .

# Independence

## Exercise 2 (Independent events)

Let  $\Omega = \{\omega_1, \omega_2, \omega_3, \omega_4\}$  equipped with the uniform probability distribution  $\mathbb{P}$ . Define the events  $A = \{\omega_1, \omega_2\}$ ,  $B = \{\omega_1, \omega_3\}$  and  $C = \{\omega_2, \omega_3\}$ . Show that A, B and C are pairwise independent. Compare  $\mathbb{P}(A \cap B \cap C)$  and  $\mathbb{P}(A)\mathbb{P}(B)\mathbb{P}(C)$ .

Solution 2 On a  $\mathbb{P}[(]\{w_i\}) = 1/|\Omega| = 1/4$  pour  $i = 1, \ldots, 4$ . D'une part,  $\mathbb{P}[(]A) = \mathbb{P}[(]\{\omega_1\}) + \mathbb{P}[(]\{\omega_2\}) = 1/2$ . De même,  $\mathbb{P}[(]B) = \mathbb{P}[(]C) = 1/2$ . D'autre part,  $A \cap B = \{\omega_1\}$  et donc  $\mathbb{P}[(]A \cap B) = 1/4$ . Donc, on a montré que  $\mathbb{P}[(]A \cap B) = \mathbb{P}[(]A)\mathbb{P}[(]B)$ , d'où l'indépendance de A et B. De même, on montre que A et C sont indépendants et B et C sont indépendants.

Comme  $A \cap B \cap C = \emptyset$ , on a  $\mathbb{P}[(]A \cap B \cap C) = 0$ . En revanche,  $\mathbb{P}[(]A)\mathbb{P}[(]B)\mathbb{P}[(]C) = 1/8$ . Cela implique que A, B et C ne sont pas mutuellement indépendants.

#### Exercise 3

Let  $A_1, \ldots, A_n$  be n events from a probability space  $(\Omega, \mathbb{P})$ . Suppose that they are mutually independent. Find an explicit expression for  $\mathbb{P}(A_1 \cup \cdots \cup A_n)$  depending on the  $\mathbb{P}(A_i)$ .

### Exercise 4

Let  $(\Omega, \mathcal{F}, \mathbb{P})$  a probability space. Let  $(A_n)_{n\geq 0}$  a series of independent events. We note  $A=\limsup_n A_n$ . Let assume that  $\sum_n \mathbb{P}(A_n)=+\infty$  and we want to prove that  $\mathbb{P}(A)=1$ .

- 1. Preliminary. Justify that for all x > -1,  $\ln(1+x) \le x$ .
- 2. Let  $n \leq N$ . We note  $E_{n,N} = \bigcap_{k=n}^N A_k^c$  and  $E_n = \bigcap_{k>n} A_k^c$ .
  - (a) Prove that (n fixed),  $\lim_{N\to+\infty} \ln (\mathbb{P}(E_{n,N})) = -\infty$ .
  - (b) Deduce that  $\mathbb{P}(E_n) = 0$ .
  - (c) Deduce that  $\mathbb{P}(A) = 1$ .

Solution 3 1. La fonction ln est concave. Sa courbe représentative est en-dessous de sa tangente au point d'abscisse 1. L'inégalité demandée est juste la traduction analytique de cette propriété géométrique.

2. (a) Les événements  $A_k$  étant indépendants, il en est de même des événements  $\overline{A_k}$ , et donc

$$P(E_{n,N}) = \prod_{k=n}^{N} P(\overline{A_k}) = \prod_{k=n}^{N} (1 - P(A_k)).$$

En utilisant l'inégalité précédente, on a

$$\ln \left( P\left( E_{n,N} \right) \right) \le - \sum_{k=n}^{N} P\left( A_{k} \right).$$

Puisque  $\sum_{k\geq n} P\left(A_k\right) = +\infty$ , on en déduit le résultat.

(b) Par composition par la fonction exponentielle,  $(P(E_{n,N}))$  tend vers 0 lorsque N tend vers l'infini (et n reste fixé). Mais, la suite  $(E_{n,N})_N$  est décroissante et

$$E_n = \bigcap_{N \ge n} E_{n,N}$$

Ainsi,

$$P(E_n) = \lim_{N} P(E_{n,N}) = 0.$$

(c) A s'écrit  $A = \bigcap_n \overline{E_n}$ . La suite  $(\overline{E_n})$  est décroissante et  $P(\overline{E_n}) = 1$ . Ainsi, on trouve que

$$P(A) = \lim_{n} P\left(\overline{E_n}\right) = 1.$$

## Random variables

#### Exercise 5

Find two random variables X and Y on a probability space  $(\Omega, \mathbb{P})$  (to be specified) having the same distribution, but that are not equal.

## Exercise 6

In an oil region, the probability that one drilling leads to an oil slick is 0.1.

1. Justify that one drilling can be modeled using a Bernoulli distribution.

- 2. We made 10 oil drillings. Let X be the number of drillings that led to an oil slick.
  - (a) Under which assumptions X can be modeled using a binomial distribution? Precise the parameters.
  - (b) Assume that X follows a binomial distribution. Compute
    - i. the probability that exactly two drillings lead to oil slicks.
    - ii. the probability that at least one drilling leads to an oil slick.

#### Exercise 7

Let  $\lambda > 0$  be fixed. Let  $X_n, n \ge 1$  be random variables with binomial distribution with parameters n and  $\lambda/n$ , and Y be a random variable with Poisson distribution with parameter  $\lambda$ . Show that, for any  $k \in \mathbb{N}$ ,

$$\lim_{n \to +\infty} \mathbb{P}(X_n = k) = \mathbb{P}(Y = k).$$

Hint: Use Striling's approximation:  $n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$ .

We will later see that this result means that  $X_n$  converges in distribution to Y, or, to put it differently, that the binomial distribution with parameters n and  $\lambda/n$  converges to the Poisson distribution with parameter  $\lambda$ .

#### **Solution 4** We have:

$$\mathbb{P}[X_n = k] = \binom{n}{k} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^{n-k}$$
$$= \frac{n!}{k!(n-k)!} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^{n-k}$$
$$= \frac{\lambda^k}{k!} \frac{n!}{(n-k)!n^k} \left(1 - \frac{\lambda}{n}\right)^{n-k}$$

And:

$$\left(1 - \frac{\lambda}{n}\right)^{n-k} = \left(1 - \frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^n$$

with

$$\left(1 - \frac{\lambda}{n}\right)^k \xrightarrow[n \to \infty]{} 0$$

and

$$\left(1 - \frac{\lambda}{n}\right)^n = e^{\log\left(1 - \frac{\lambda}{n}\right)^n}$$

$$= e^{n \cdot \log\left(1 - \frac{\lambda}{n}\right)}$$

$$\sim e^{n\left(-\frac{\lambda}{n}\right)}$$

$$= e^{-\lambda}$$

Also,

$$\frac{n!}{(n-k)!} = n.(n-1).(n-2).\cdot.(n-k+1) \sim n^k$$

Finally,

$$\binom{n}{k} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^{n-k} \sim \frac{\lambda^k}{k!} 1.e^{-\lambda}$$

This conclude the exercise.

# Expectation

### Exercise 8

Compute the mean, variance and cumulated distribution function of

- 1. the binomial distribution Bin(n, p) with  $n \ge 1$  and p > 0.
- 2. the Poisson distribution  $Poi(\lambda)$  with  $\lambda > 0$ .
- 3. the uniform distribution U[a, b] with a < b.
- 4. the exponential distribution  $\text{Exp}(\lambda)$  with  $\lambda > 0$ .
- 5. the normal distribution  $\mathcal{N}(\mu, \sigma^2)$  of probability density function

$$f(x): x \to \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

with  $\mu \in \mathbb{R}$  and  $\sigma > 0$ .

## Exercise 9

- 1. Show that if X exponential distribution  $\text{Exp}(\lambda)$  with  $\lambda > 0$ , then  $\mathbb{E}[X^n] = \frac{n!}{\lambda^n}$ ;
- 2. Show that if X follows  $\mathcal{N}(0,1)$  then  $\mathbb{E}\left[X^{2n}\right] = \prod_{k=1}^{n} (2k-1) = \frac{(2n)!}{2^n n!}$

Solution 5 1. Par intégration par parties :

$$\mathbb{E}\left(X^{n}\right)=\int_{0}^{+\infty}x^{n}\lambda e^{-\lambda x}dx=\int_{0}^{+\infty}x^{n-1}e^{-\lambda x}dx=\frac{n}{\lambda}\mathbb{E}\left(X^{n-1}\right).$$

On en déduit le résultat par récurence immédiate.

2. Par intégration par parties:

$$\mathbb{E}\left(X^{2n}\right) = \int_{\mathbb{R}} x^{2n} \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} dx = \int_{\mathbb{R}} \frac{x^{2n+2}}{2n+1} \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} dx = \frac{1}{2n+1} \mathbb{E}\left(X^{2(n+1)}\right).$$

On en déduit le résultat par récurence immédiate.

### Exercise 10

Let  $X: \Omega \to [0; +\infty]$  (note that  $+\infty$  is allowed) be a random variable such that  $\mathbb{E}[X] < \infty$ .

- 1. Prove that X is finite almost surely (proceed by contradiction).
- 2. Assume that  $\mathbb{E}[X] = 0$ . Prove that X = 0 almost surely. Hint: use that  $X \geq \frac{1}{n} \mathbf{1}_{X \geq 1/n}$ .

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# Variance Inequalities

### Exercise 11

Let X be a random variable such that  $\mathbb{E}\left[X^2\right]<+\infty$ . Prove that :

- 1.  $0 \leq \operatorname{Var}(X) < \infty$
- 2.  $\operatorname{Var}(X) = \mathbb{E}\left[X^2\right] (\mathbb{E}[X])^2$ .
- 3.  $Var(X) = 0 \iff \mathbb{P}(X = c) = 1$  for some constant c.
- 4. For any constants  $a, b, Var(aX + b) = Var(aX) = a^2 Var(X)$ .

#### **Solution 6** 1. We will show here:

- (a)  $\mathbb{E}[X]$  exists and is finite.
- (b)  $Var(X) \geq 0$
- (c)  $Var(X) < \infty$

Let begin.

- (a)  $\mathbb{E}[X^2]$  is finite so  $\mathbb{E}[(X+1)^2]$  is. So  $\mathbb{E}[X^2+2X+1]$  is. As  $\mathbb{E}[X^2]+1$  is obviously finite,  $\mathbb{E}[X]$  exists and is finite.
- (b) We have  $\operatorname{Var}[X] = \mathbb{E}[(X \mathbb{E}[X])^2]$  which exists as  $\mathbb{E}[X]$  is a constant. As  $(X \mathbb{E}[X])^2$  is almost surely positive, we have the wanted result.
- (c) We have:

$$\begin{aligned} \operatorname{Var}[X] &= \mathbb{E}[(X - \mathbb{E}[X])^2] \\ &= \mathbb{E}[X^2] + \mathbb{E}[-2.X.\mathbb{E}[X]] + \mathbb{E}[X]^2 \\ &= \mathbb{E}[X^2] - 2.\mathbb{E}[X].\mathbb{E}[X] + \mathbb{E}[X]^2 \\ &= \mathbb{E}[X^2] - \mathbb{E}[X]^2 \end{aligned}$$

The two term are finite. Therefore, Var[X] is finite.

- 2. Following the previous calculus, the wanted equality is true.
- 3. We have:

$$\begin{aligned} \operatorname{Var}[X] &= 0 \iff \operatorname{Var}[(X - \operatorname{\mathbb{E}}[X])^2] = 0 \\ &\iff X - \operatorname{\mathbb{E}}[X] = 0 \\ almost surely \\ &\iff \exists c \in \mathbb{R}, \operatorname{\mathbb{P}}[X = c] = 1 \end{aligned}$$

4. We have :

$$Var[aX + b] = \mathbb{E}[(aX + b - \mathbb{E}[aX + b])^{2}]$$

$$= \mathbb{E}[(aX - \mathbb{E}[ax])^{2}] = Var[aX]$$

$$= a^{2}\mathbb{E}[(X - \mathbb{E}[X])^{2}]$$

$$= a^{2}Var[X]$$

## Exercise 12

Let X be non-negative  $(X \ge 0 \text{ a.s.})$  and a > 0 be a constant.

1. Justify that

$$\forall \omega \in \Omega, \quad a\mathbf{1}_{\{X(\omega) \geqslant a\}} \leqslant X(\omega)\mathbf{1}_{\{X(\omega) \geqslant a\}} \leqslant X(\omega)$$

2. Prove the Markov's inequality

$$\mathbb{P}(X \ge a) \le \frac{\mathbb{E}[X]}{a}$$

Solution 7 1. Let prove that:

(a)  $\forall \omega \in \Omega$ ,  $a\mathbf{1}_{\{X(\omega)\geqslant a\}} \leqslant X(\omega)\mathbf{1}_{\{X(\omega)\geqslant a\}}$ 

(b) 
$$\forall \omega \in \Omega$$
,  $X(\omega)\mathbf{1}_{\{X(\omega)\geqslant a\}} \leqslant X(\omega)$ 

Let us begin.

- (a) Let  $\omega \in \Omega$ . Consider the two cases  $X(\omega) \geq a$  and  $X(\omega) < a$  and prove that the wanted inequality is true in the two cases.
- (b) As the indicator function is below or equal to 1, the inequality is true.
- 2. We have from 1. :

$$\forall \omega \in \Omega, \quad a\mathbf{1}_{\{X(\omega) \geqslant a\}} \leqslant X(\omega)$$

Taking the expected value in this inequality:

$$\mathbb{E}[a\mathbf{1}_{\{X(\omega)\geqslant a\}}] \leq \mathbb{E}[X]$$

$$\iff a\mathbb{E}[\mathbf{1}_{\{X(\omega)\geqslant a\}}] \leq \mathbb{E}[X]$$

$$\iff a\mathbb{P}[X \geq a] \leq \mathbb{E}[X]$$

$$\iff \mathbb{P}[X \geq a] \leq \frac{\mathbb{E}[X]}{a}$$

## Exercise 13

Assume that  $\mathbb{E}\left[X^2\right]<+\infty$ . Applying Markov's inequality to  $(X-\mathbb{E}[X])^2$  prove that, for any constant a>0,

$$\mathbb{P}(|X - \mathbb{E}[X]| \ge a) \le \frac{\operatorname{Var}(X)}{a^2}$$

**Solution 8** Applying markov inequality to  $(X - \mathbb{E}[X])^2$  and  $a^2 > 0$ , we have :

$$\mathbb{P}[(X - \mathbb{E}[X])^2 \ge a^2] \le \frac{\mathbb{E}[(X - \mathbb{E}[X])^2]}{a^2}$$
$$\iff \mathbb{P}[|X - \mathbb{E}[X]| \ge a] \le \frac{\operatorname{Var}[X]}{a^2}$$