Advanced Probability and Applications

### Solutions to Homework 2

# Exercise 1.

a) 1. true, 2. true, 3. false, 4. true b) 5. false, 6. true, 7. false, 8. true.

# Exercise 2.

a) Use  $B = A \cup (B \setminus A)$ , where A and  $B \setminus A$  are disjoint, as well as  $\Omega = A \cup A^c$  and  $\mathbb{P}(\Omega) = 1$ .

b) Use  $A \cup B = A \cup (B \setminus (A \cap B))$  where A and  $B \setminus (A \cap B)$  are disjoint, as well as a).

c) Use  $\bigcup_{n=1}^{\infty} A_n = \bigcup_{n=1}^{\infty} B_n$ , where  $B_n = A_n \setminus (A_1 \cup \ldots \cup A_{n-1})$ ; the  $B_n$  are disjoint, so by axiom (ii) and a),

$$\mathbb{P}(\bigcup_{n=1}^{\infty} A_n) = \mathbb{P}(\bigcup_{n=1}^{\infty} B_n) = \sum_{n=1}^{\infty} \mathbb{P}(B_n) \le \sum_{n=1}^{\infty} \mathbb{P}(A_n)$$

d)  $\mathbb{P}(\bigcup_{n\geq 1}A_n) = \mathbb{P}(\bigcup_{n\geq 1}(A_n\cap A_{n-1}^c)) \stackrel{(*)}{=} \sum_{n=1}^{\infty} \mathbb{P}(A_n\cap A_{n-1}^c) = \lim_{n\to\infty} \sum_{i=1}^n \mathbb{P}(A_i\cap A_{i-1}^c))$  $\stackrel{(**)}{=} \lim_{n\to\infty} \mathbb{P}(\bigcup_{i=1}^n (A_i\cap A_{i-1}^c)) = \lim_{n\to\infty} \mathbb{P}(\bigcup_{i=1}^n A_i) = \lim_{n\to\infty} \mathbb{P}(A_n)$ , where (\*), (\*\*) follow from the fact that the sets  $A_n \cap A_{n-1}^c$  are disjoint.

e) Using parts a) and d):  $\mathbb{P}(\bigcap_{n\geq 1}A_n) = 1 - \mathbb{P}((\bigcap_{n\geq 1}A_n)^c) = 1 - \mathbb{P}(\bigcup_{n\geq 1}A_n^c) = 1 - \lim_{n\to\infty}\mathbb{P}(A_n^c) = \lim_{n\to\infty}\mathbb{P}(A_n).$ 

# Exercise 3.\*

a) First, note that the range of the random variable X is [0, 1]. Thus, the CDF  $F_X(t) = 0$  for t < 0and  $F_Y(t) = 1$  for  $t \ge 1$ .

Now, for  $t \in [0, 1]$ , we have:

$$F_X(t) = \mu_X((-\infty, t]) = \mathbb{P}(\{(\omega_1, \omega_2) \in [0, 1]^2 : X(\omega_1, \omega_2) \le t\})$$
  
=  $\mathbb{P}(\{(\omega_1, \omega_2) \in [0, 1]^2 : \omega_1 \omega_2 \le t\})$ 

Now, one could just compute the probability by integrating the area under the curve  $\omega_1 \omega_2 \leq t$  that lies within  $[0,1] \times [0,1]$  as follows:

$$\mathbb{P}(\{(\omega_1, \omega_2) \in [0, 1]^2 : \omega_1 \omega_2 \le t\}) = t + \int_t^1 \frac{t}{\omega_1} d\omega_1$$
$$= t(1 - \ln t)$$

Therefore,

$$F_X(t) = \begin{cases} 0 & \text{if } t < 0, \\ t(1 - \ln t) & \text{if } t \in [0, 1] \\ 1 & \text{if } t > 1 \end{cases}$$

b) First, note that the range of the random variable Y is  $\left[-\frac{1}{2},\frac{1}{2}\right]$ . Thus, the CDF  $F_Y(t) = 0$  for  $t < -\frac{1}{2}$  and  $F_Y(t) = 1$  for  $t \ge \frac{1}{2}$ .

Now, for  $t \in \left[-\frac{1}{2}, \frac{1}{2}\right]$ , we have:

$$F_Y(t) = \mu_Y((-\infty, t]) = \mathbb{P}(\{(\omega_1, \omega_2) \in [0, 1]^2 : Y(\omega_1, \omega_2) \le t\})$$
  
=  $\mathbb{P}(\{(\omega_1, \omega_2) \in [0, 1]^2 : \omega_1 - \omega_2 \le 2t\})$ 

Note that the area  $\{(\omega_1, \omega_2) \in [0, 1]^2 : \omega_1 - \omega_2 \leq 2t\}$  represents different shapes in  $[0, 1] \times [0, 1]$  for positive and negative values of 2t. Thus, we divide our analysis into two cases:

**Case 1:**  $-\frac{1}{2} < t \le 0$ :

The area  $\{(\omega_1, \omega_2) \in [0, 1]^2 : \omega_1 - \omega_2 \leq 2t\}$  represents a right-angled triangle  $(\Delta_1)$  is an element of the sigma field  $\mathcal{F} = \mathcal{B}([0, 1]^2)$ . Thus, the probability measure  $\mathbb{P}(\Delta_1)$  is given by its area. Thus,

$$F_Y(t) = Area(\Delta_1) = \frac{1}{2}(1+2t)(1+2t)$$

**Case 2:**  $0 < t \le \frac{1}{2}$ :

The area  $\{(\omega_1, \omega_2) \in [0, 1]^2 : \omega_1 - \omega_2 \leq 2t\}$  represents a pentagon  $(\Delta_2)$  in this case which is again an element of the sigma field  $\mathcal{F} = \mathcal{B}([0, 1]^2)$ . Thus, the probability measure  $\mathbb{P}(\Delta_2)$  is given by its area which can be easily computed as:

$$F_Y(t) = Area(\Delta_2) = 1 - \frac{1}{2}(1 - 2t)(1 - 2t)$$

Thus, the CDF of the random variable Y is the following:

$$F_Y(t) = \begin{cases} 0 & \text{if } t \le -\frac{1}{2}, \\ \frac{1}{2}(1+2t)^2 & \text{if } -\frac{1}{2} < t \le 0\\ 1 - \frac{1}{2}(1-2t)^2 & \text{if } 0 < t \le \frac{1}{2}\\ 1 & \text{if } t > \frac{1}{2} \end{cases}$$

Exercise 4.

a) We have

$$\mathbb{P}(\{Y_n \le t\}) = 1 - \mathbb{P}(\{Y_n > t\}) = 1 - \mathbb{P}(\{\min\{X_1, \dots, X_n\} > t\}) = 1 - \mathbb{P}(\bigcap_{j=1}^n \{X_j > t\}) = 1 - \prod_{j=1}^n \mathbb{P}(\{X_j > t\}) = 1 - \mathbb{P}(\{X_1 > t\})^n$$

where the last two equalities follow from the assumption that the X's are i.i.d. Therefore,

$$\mathbb{P}(\{Y_n \le t\}) = 1 - (\exp(-t))^n = 1 - \exp(-nt)$$

b) Under the assumptions made, n is large and t is such that  $nt \ll 1$ , so using Taylor's expansion  $\exp(-x) \simeq 1 - x$ , we obtain

$$\mathbb{P}(\{Y_n \le t\}) \simeq 1 - (1 - nt) = nt \quad \text{while} \quad \mathbb{P}(\{X_1 \le t\}) = 1 - \exp(-t) \simeq t$$

and therefore  $\mathbb{P}(\{Y_n \leq t\}) \simeq n \mathbb{P}(\{X_1 \leq t\}).$ 

c) We have similarly

$$\mathbb{P}(\{Z_n \ge t\}) = 1 - \mathbb{P}(\{Z_n < t\}) = 1 - \mathbb{P}(\{\max\{X_1, \dots, X_n\} < t\}) = 1 - \mathbb{P}(\cap_{j=1}^n \{X_j < t\})$$
$$= 1 - \prod_{j=1}^n \mathbb{P}(\{X_j < t\}) = 1 - \mathbb{P}(\{X_1 < t\})^n = 1 - (1 - \exp(-t))^n$$

d) Under the assumptions made, n is large and t is such that  $n \exp(-t) \ll 1$ , so using again the same Taylor expansion as above, we obtain

$$\mathbb{P}(\{Z_n \ge t\}) \simeq 1 - (1 - n \exp(-t)) = n \exp(-t) \quad \text{while} \quad \mathbb{P}(\{X_1 \ge t\}) = \exp(-t)$$

and therefore  $\mathbb{P}(\{Z_n \ge t\}) \simeq n \mathbb{P}(\{X_1 \ge t\}).$ 

#### Exercise 5.

a) Yes. Here, we need to check that  $\forall B \in \mathcal{B}(\mathbb{R})$ , we have  $X^{-1}(B) \in \mathcal{F}$ . Specifically, check for  $B = \{0\}, \{-2\}, \{1\}, \{0, 1\}, \{0, -2\}, \{1, -2\}.$ 

b) No. Here, we have  $X^{-1}(-1) = \{a, b\}, X^{-1}(1) = \{c\}, X^{-1}(2) = \{d\}$ . However,  $\{c\}, \{d\} \notin A$ . Thus, X is not a  $\mathcal{F}$ -measurable random variable.

c) Let us begin by recalling that the cdf is a non-decreasing and a right-continuous function i.e.,  $\forall m \in \mathbb{R}$ , we have:

$$\lim_{\epsilon \to 0} F(m-\epsilon) \le F(m) = \lim_{\epsilon \to 0} F(m+\epsilon)$$

Now, defining  $m_0 := \sup\{x \in \mathbb{R} : F(x) < 1/2\}$ . Then,  $F(m_0) \ge 1/2$ . How??. Assume, it's not true i.e.,  $F(m_0) < 1/2$ . Since, the cdf F is a right continuous function, it implies that  $\exists m'_0 > m_0$  such that  $F(m'_0) < 1/2$ . However, this violates our definition of  $m_0$ . Thus, our assumption is wrong and consequently  $F(m_0) \ge 1/2$ . Furthermore,  $\lim_{\epsilon \to 0} F(m_0 - \epsilon) \le 1/2$ . Hence,  $m_0$  is the (smallest) median of X.

Let's also define  $m_1 := \sup\{x \in \mathbb{R} : F(x) \le 1/2\}$ . Then,  $F(m_1) \ge 1/2$  as F is a non-decreasing function. Furthermore,  $\lim_{\epsilon \to 0} F(m_1 - \epsilon) \le 1/2$ . Thus,  $m_1$  is the (largest) median of X.

Now, note that  $\{x \in \mathbb{R} : F(x) < 1/2\} \subseteq \{x \in \mathbb{R} : F(x) \le 1/2\}$ . Thus,  $m_0 \le m_1$ . We can now show that the closed interval  $[m_0, m_1]$  contains the medians i.e.,  $\forall m \in [m_0, m_1]$ , m is a median of X. Since,  $m_0 \le m \le m_1$ , it's not hard to see that  $1/2 \le F(m_0) \le F(m) \le F(m_1)$ . Furthermore, note that  $\forall \epsilon > 0$ ,  $\lim_{\epsilon \to 0} F(m_0 - \epsilon) \le \lim_{\epsilon \to 0} F(m - \epsilon) \le \lim_{\epsilon \to 0} F(m_1 - \epsilon) \le 1/2$ . Thus, we have that  $\forall m \in [m_0, m_1]$ ,  $\lim_{\epsilon \to 0} F(m - \epsilon) \le \frac{1}{2} \le F(m)$ .

The median of X is unique when the cdf of X is a strictly increasing function. In this case, the closed interval  $[m_0, m_1]$  reduces to a singleton set.