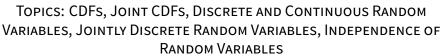
Al 5030: Probability and Stochastic Processes

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HOMEWORK 5





Fix a probability space $(\Omega, \mathscr{F}, \mathbb{P})$. All random variables appearing below are assumed to be defined with respect to \mathscr{F} .

1. Let X be a random variable. Determine, in each case below, if the function therein can be a valid CDF of X. If not, provide at least one valid justification. For each valid CDF, compute $\mathbb{P}(\{X > 5\})$.

(a)
$$F_X(x) = \begin{cases} \frac{e^{-x^2}}{4}, & x < 0, \\ 1 - \frac{e^{-x^2}}{4}, & x \ge 0. \end{cases}$$

Solution: A valid CDF. Follows right continuity, $\lim_{x\to\infty}F_X(x)=1$ and $\lim_{x\to-\infty}F_X(x)=0$ and the function is monotonically increasing.

$$\mathbb{P}(\{X > 5\}) = 1 - \mathbb{P}(\{X \le 5\}) = 1 - F_X(5) = \frac{e^{-25}}{4}.$$

(b)
$$F_X(x) = \begin{cases} 0, & x < 0, \\ 0.5 + e^{-x}, & 0 \le x < 3, \\ 1, & x \ge 3. \end{cases}$$

Solution: Not a valid CDF. Follows right continuity, $\lim_{x\to\infty} F_X(x)=1$ and $\lim_{x\to-\infty} F_X(x)=0$; but the function is monotonically decreasing when $0\leq x<3$. We can also see that $\mathbb{P}(\{X=0\})+\mathbb{P}(\{X=3\})>1$.

(c)
$$F_X(x) = \begin{cases} 0, & x < 0, \\ 0.5 + \frac{x}{20}, & 0 \le x < 10, \\ 1, & x \ge 10. \end{cases}$$

Solution: A valid CDF. Follows right continuity, $\lim_{x\to\infty} F_X(x)=1$ and $\lim_{x\to-\infty} F_X(x)=0$ and the function is monotonically increasing.

$$\mathbb{P}(\{X > 5\}) = 1 - \mathbb{P}(\{X \le 5\}) = 1 - F_X(5) = 1 - 0.5 - \frac{5}{20} = 0.25.$$

2. Let X be a Geometric random variable. Show that for all $n,k\geq 1$,

$$\mathbb{P}(\{X > n + k\} | \{X > n\}) = \mathbb{P}(\{X > k\}).$$

This is called the memoryless property of a Geometric distribution.

Conversely, show that if the random variable X satisfies the above property, then it must be Geometric.

Solution: Assuming X is a Geometric RV with parameter (probability of success) p. We have that $\mathbb{P}(\{X>n\})=\sum_{r=n}^{\infty}(1-p)^rp=(1-p)^n$. Then,

$$\mathbb{P}(\{X > n + k\} | \{X > n\}) = \frac{\mathbb{P}(\{X > n + k\} \cap \{X > n\})}{\mathbb{P}(\{X > n\})} \stackrel{(a)}{=} \frac{\mathbb{P}(\{X > n + k\})}{\mathbb{P}(\{X > n\})}$$

$$= \frac{(1 - p)^{n + k}}{(1 - p)^n}$$

$$= (1 - p)^k$$

$$= \mathbb{P}(\{X > k\}),$$

where (a) above follows by noting that for any $\omega \in \Omega$ such that $X(\omega) > n + k$, we have $X(\omega) > n$. Therefore, $\{X > n + k\} \subseteq \{X > n\}$, from which it follows that $\{X > n + k\} \cap \{X > n\} = \{X > n + k\}$.

Conversely, assume that the relation $\mathbb{P}(\{X>n+k\}|\{X>n\})=\mathbb{P}(\{X>k\})$ holds for all $n,k\geq 1$. In particular, note that it holds for k=1. This gives us $\mathbb{P}(\{X>n+1\}|\{X>n\})=\mathbb{P}(\{X>1\})\Rightarrow \frac{\mathbb{P}(\{X>n+1\})}{\mathbb{P}(\{X>n\})}=\mathbb{P}(\{X>1\})$. Applying recursively, we get $\mathbb{P}(\{X>n\})=\mathbb{P}(\{X>1\})^n$. Defining $p:=1-\mathbb{P}(X>1)$, we have $\mathbb{P}(\{X>n\})=(1-p)^n$, which matches the CDF of the Geometric distribution.

3. Let X and Y be continuous random variables with PDFs f_X and f_Y respectively. For any $\alpha \in [0,1]$, argue that $\alpha f_X + (1-\alpha) f_Y$ is a valid PDF. Can you think of a random variable Z whose PDF is $f_Z = \alpha f_X + (1-\alpha) f_Y$?

Solution: Clearly, the function $f(x) = \alpha f_X(x) + (1 - \alpha) f_Y(x)$, $x \in \mathbb{R}$, is non-negative. Furthermore,

$$\int_{-\infty}^{\infty} f(x) dx = \alpha \int_{-\infty}^{\infty} f_X(x) dx + (1 - \alpha) \int_{-\infty}^{\infty} f_Y(x) dx = 1,$$

where the last equality follows by noting that $\int_{-\infty}^{\infty} f_X(x) \, dx = 1 = \int_{-\infty}^{\infty} f_Y(x) \, dx$, as f_X and f_Y are valid PDFs. Let $B \sim \text{Ber}(\alpha)$ be a Bernoulli random variable independent of X and Y, i.e., $B \perp X$ and $B \perp Y$. Let Z be a random variable defined as

$$Z(\omega) = \begin{cases} X(\omega), & B(\omega) = 1, \\ Y(\omega), & B(\omega) = 0, \end{cases} \quad \omega \in \Omega.$$

The CDF of Z is given by

$$F_{Z}(z) = \mathbb{P}(\{Z \le z\})$$

$$\stackrel{(a)}{=} \mathbb{P}(\{Z \le z\} \cap \{B = 1\}) + \mathbb{P}(\{Z \le z\} \cap \{B = 0\})$$

$$\stackrel{(b)}{=} \mathbb{P}(\{X \le z\} \cap \{B = 1\}) + \mathbb{P}(\{Y \le z\} \cap \{B = 0\})$$

$$\stackrel{(c)}{=} \mathbb{P}(\{X \le z\}) \cdot \mathbb{P}(\{B = 1\}) + \mathbb{P}(\{Y \le z\}) \cdot \mathbb{P}(\{B = 0\})$$

$$= \alpha F_{X}(z) + (1 - \alpha) F_{Y}(z), \qquad z \in \mathbb{R},$$

where (a) above follows from the law of total probability, (b) above follows from the definition of Z, and (c) above follows from the fact that $B \perp X$ and $B \perp Y$. Differentiating the CDF of Z, we obtain the PDF of Z as

$$f_Z(z) = \alpha f_X(z) + (1 - \alpha) f_Y(z), \qquad z \in \mathbb{R}.$$

4. We learnt in class that random variables X and Y are independent if

$$\{X \le x\} \perp \{Y \le y\} \qquad \forall x, y \in \mathbb{R}. \tag{1}$$

In this exercise, we will show through a series of logical steps that the above definition of independence implies that

$$\{X = x\} \perp \!\!\!\perp \{Y = y\} \qquad \forall x, y \in \mathbb{R}. \tag{2}$$

Fix arbitrary $x, y \in \mathbb{R}$.

- (a) Show that if A, B are events such that $A \perp \!\!\! \perp B$, then $A \perp \!\!\! \perp B^c$.
- (b) Show that if A, B, C are events such that $A \perp \!\!\! \perp B$ and $A \perp \!\!\! \perp C$, then $A \perp \!\!\! \perp (B \cup C)$ if and only if $A \perp \!\!\! \perp (B \cap C)$.
- (c) Let $B_1, B_2, \ldots \in \mathscr{F}$ be such that $B_n \supseteq B_{n+1}$ for all $n \in \mathbb{N}$. If $A \perp \!\!\! \perp B_n$ for all $n \in \mathbb{N}$, show that $A \perp \!\!\! \perp \bigcap_{n=1}^\infty B_n$.
- (d) Apply the result in part (c) to the sets $A = \{X \le x\}$ and $B_n = \{Y > y \frac{1}{n}\}$. Argue using the result in part (a) that $A \perp \!\!\!\perp B_n$ for all n. Hence conclude from the result in part (c) that $\{X \le x\} \perp \!\!\!\perp \{Y \ge y\}$.
- (e) Apply the result in part (b) to the sets $A=\{X\leq x\}$, $B=\{Y\leq y\}$, and $C=\{Y\geq y\}$, and conclude using the results in parts (d) and (a) that $\{X\leq x\}\perp\{Y=y\}$. What are $B\cap C$ and $B\cup C$ here?
- (f) Repeat the series of logical steps in parts (d) and (e) with $A = \{Y = y\}$ to arrive at $\{X = x\} \perp \{Y = y\}$.

Remark: This exercise shows that if X and Y are independent, then (1) implies (2). However, in general, (2) DOES NOT imply (1). If X and Y are jointly discrete, then (2) implies (1).

Solution: We prove each of the parts below in order.

(a) We have $A=(A\cap B^c)\cup (A\cap B)$, and if $A\perp B$, then $\mathbb{P}(A\cap B)=\mathbb{P}(A)$ $\mathbb{P}(B)$. We then note that

$$\mathbb{P}(A) = \mathbb{P}((A \cap B^c) \cup (A \cap B))$$

$$\mathbb{P}(A) = \mathbb{P}(A \cap B^c) + \mathbb{P}(A \cap B)$$

$$\mathbb{P}(A) = \mathbb{P}(A \cap B^c) + \mathbb{P}(A) \mathbb{P}(B),$$

from which it follows after re-arranging terms that $\mathbb{P}(A \cap B^c) = \mathbb{P}(A) \mathbb{P}(B^c)$, thereby proving that $A \perp \!\!\! \perp B^c$.

- (b) See the solution to question 6 of homework 4.
- (c) Because $A \perp \!\!\! \perp B_n$ for all $n \in \mathbb{N}$, we have $\mathbb{P}(A \cap B_n) = \mathbb{P}(A) \mathbb{P}(B_n)$ for all $n \in \mathbb{N}$. Taking limits as $n \to \infty$, we then get $\lim_{n \to \infty} \mathbb{P}(A \cap B_n) = \lim_{n \to \infty} \mathbb{P}(A) \mathbb{P}(B_n) = \mathbb{P}(A) \lim_{n \to \infty} \mathbb{P}(B_n)$. Because $B_n \supseteq B_{n+1}$ for all $n \in \mathbb{N}$, we note that $\lim_{n \to \infty} B_n = \bigcap_{n=1}^\infty B_n$. Furthermore, we have $A \cap B_n \supseteq A \cap B_{n+1}$ for all $n \in \mathbb{N}$, and therefore $\lim_{n \to \infty} A \cap B_n = \bigcap_{n=1}^\infty A \cap B_n = A \cap \bigcap_{n=1}^\infty B_n$. Applying continuity of probability, we get

$$\begin{split} &\lim_{n \to \infty} \mathbb{P}(A \cap B_n) = \lim_{n \to \infty} \mathbb{P}(A) \, \mathbb{P}(B_n) \\ \mathbb{P}\big(\lim_{n \to \infty} (A \cap B_n)\big) = \mathbb{P}(A) \, \mathbb{P}(\lim_{n \to \infty} B_n) \\ \mathbb{P}(A \cap \lim_{n \to \infty} B_n) = \mathbb{P}(A) \, \mathbb{P}(\lim_{n \to \infty} B_n) \\ \mathbb{P}\left(A \cap \left(\bigcap_{n=1}^{\infty} B_n\right)\right) = \mathbb{P}(A) \, \mathbb{P}\left(\bigcap_{n=1}^{\infty} B_n\right). \end{split}$$

We have thus proved that $A \perp \bigcap_{n=1}^{\infty} B_n$.

- (d) It is clear that, $B_1, B_2, \ldots \in \mathscr{F} \& B_n \supseteq B_{n+1}$ for all $n \in \mathbb{N}$. Then, $\lim_{n \to \infty} B_n = \bigcap_{n=1}^\infty B_n = \{Y \ge y\}$. Let us consider $C_n = B_n^c = \{Y \le y \frac{1}{n}\}$. Because X and Y are independent, it follows that $A \perp C_n$ for all $n \in \mathbb{N}$, where A is as defined in the question. Using the result in part (a), it follows that $A \perp C_n^c = B_n$ for all $n \in \mathbb{N}$. Applying the result in part (c), we get that $A \perp \bigcap_{n=1}^\infty B_n = \{Y \ge y\}$.
- (e) Here, $A = \{X \leq x\}, B = \{Y \leq y\}$, and $C = \{Y \geq y\}$. Then, $B \cup C = \{Y \in \mathbb{R}\} = \Omega$, and $B \cap C = \{Y = y\}$.

$$\begin{split} \mathbb{P}(A \cap (B \cup C)) &= \mathbb{P}(A \cap \Omega) \\ \mathbb{P}(A \cap \Omega) &= \mathbb{P}(A) \\ \mathbb{P}(A \cap \Omega) &= \mathbb{P}(A) \mathbb{P}(\Omega). \end{split}$$

Thus $A \perp \!\!\! \perp (B \cup C)$, and from the question we know that $A \perp \!\!\! \perp B$, $A \perp \!\!\! \perp C$. Therefore, from the result in part (b), we have $A \perp \!\!\! \perp (B \cap C)$, i.e., $\{X \geq x\} \perp \!\!\! \perp \{Y = y\}$.

- (f) This follows by following along the lines of the steps presented in the solutions to parts (d) and (e) above.
- 5. Suppose that two batteries are chosen simultaneously and uniformly at random from the following group of 12 batteries: 3 new, 4 used (yet working), 5 defective. You may assume that all batteries within a particular group are identical. Let X be the number of new batteries chosen, and let Y be the number of used batteries chosen. Determine the joint PMF of X and Y, and compute $\mathbb{P}(\{|X-Y|\leq 1\})$.

Solution: Note that $X + Y \leq 2$ with probability 1. We then have

$$\begin{cases} \frac{\binom{5}{2}}{\binom{12}{2}}, & x = 0, y = 0, \\ \frac{\binom{4}{1} \cdot \binom{5}{1}}{\binom{12}{2}}, & x = 0, y = 1, \\ \frac{\binom{4}{2}}{\binom{12}{2}}, & x = 0, y = 2, \\ \frac{\binom{3}{1} \cdot \binom{5}{1}}{\binom{12}{2}}, & x = 1, y = 0, \\ \frac{\binom{3}{1} \cdot \binom{4}{1}}{\binom{12}{2}}, & x = 1, y = 1, \\ \frac{\binom{3}{2}}{\binom{12}{2}}, & x = 2, y = 0, \\ 0, & \text{otherwise.} \end{cases}$$

From the above joint PMF, it follows that

$$\mathbb{P}(\{|X - Y| \le 1\}) = p_{X,Y}(0,0) + p_{X,Y}(0,1) + p_{X,Y}(1,0) + p_{X,Y}(1,1) = \frac{57}{66}$$

6. Let $X \sim \mathsf{Uniform}([0,1])$ and Y = 1 - X. Derive the joint CDF of X and Y.

Solution: Note that

$$F_{X,Y}(x,y) = \mathbb{P}(\{X \le x\} \cap \{Y \le y\})$$

= $\mathbb{P}(\{X \le x\} \cap \{1 - X \le y\})$
= $\mathbb{P}(\{1 - y \le X \le x\}).$

Clearly, if 1-y>x (equivalently, x+y<1), then $\{1-y\leq X\leq x\}=\emptyset$, and hence $F_{X,Y}(x,y)=0$. Therefore, it suffices to characterise the joint CDF of X and Y at all points (x,y) such that $1-y\leq x$, in which case we have

$$F_{X,Y}(x,y) = \mathbb{P}(\{1 - y \le X \le x\}) = \mathbb{P}(\{1 - y < X \le x\}) = F_X(x) - F_X(1 - y).$$

Observe that

$$F_X(x) = \begin{cases} 0, & x < 0, \\ x, & 0 \le x < 1, \\ 1, & x \ge 1, \end{cases} \qquad F_X(1-y) = \begin{cases} 0, & 1-y < 0 \text{ (or } y > 1), \\ 1-y, & 0 \le 1-y < 1 \text{ (or } 0 < y \le 1), \\ 1, & 1-y \ge 1 \text{ (or } y \le 0). \end{cases}$$

We characterise the joint CDF on a case-by-case basis as below.

(a) Case 1: x < 0, y > 1, $1 - y \le x$. In this case, we have

$$F_{XY}(x,y) = F_X(x) - F_X(1-y) = 0 - 0 = 0.$$

(b) Case 2: $0 \le x < 1, y > 1, 1 - y \le x$.

In this case, we have

$$F_{X,Y}(x,y) = F_X(x) - F_X(1-y) = x - 0 = x.$$

(c) Case 3: $0 \le x < 1, 0 < y \le 1, 1 - y \le x$. In this case,

$$F_{XY}(x,y) = F_X(x) - F_X(1-y) = x - (1-y) = x + y - 1.$$

(d) Case 4: $x \ge 1, y > 1, 1 - y \le x$. In this case.

$$F_{XY}(x,y) = F_X(x) - F_X(1-y) = 1 - 0 = 1.$$

(e) Case 5: $x \ge 1$, $0 < y \le 1$, $1 - y \le x$.

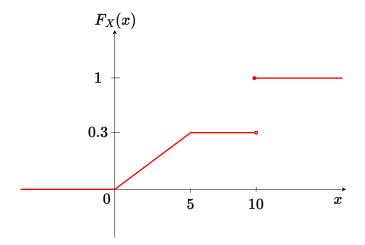
In this case,

$$F_{X,Y}(x,y) = F_X(x) - F_X(1-y) = 1 - (1-y) = y.$$

(f) Case 6: $x \ge 1$, $y \le 0$, $1 - y \le x$. In this case,

$$F_{X,Y}(x,y) = F_X(x) - F_X(1-y) = 1 - 1 = 0.$$

7. Let X be a random variable with CDF F_X as shown in the figure below. Compute $\mathbb{P}(\{X \in [3,15]\} | \{X > 4\})$.



Solution: We have

$$\mathbb{P}(\{X \in [3, 15]\} | \{X > 4\}) = \frac{\mathbb{P}(\{X \in [3, 15]\} \cap \{X > 4\})}{\mathbb{P}(\{X > 4\})}$$

$$= \frac{\mathbb{P}(\{X \in (4, 15]\})}{1 - \mathbb{P}(\{X \le 4\})}$$

$$= \frac{F_X(15) - F_X(4)}{1 - F_X(4)}$$

$$= \frac{1 - F_X(4)}{1 - F_X(4)}$$

$$= 1,$$

where the penultimate line follows by noting that $F_X(15) = 1$.