

Chapter 2: Random Variables

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- 2.** Let X be such that $P(X = 2) = P(X = 3) = 1/10$ and $P(X = 5) = 8/10$. Find the CDF and evaluate $P(2 < X \leq 4.8)$ and $P(2 \leq X \leq 4.8)$.

Solution. By definition,

$$F(x) = P(X \leq x) = \begin{cases} 0 & \text{if } x < 2, \\ 1/10 & \text{if } 2 \leq x < 3, \\ 2/10 & \text{if } 3 \leq x < 5, \\ 1 & \text{if } x \geq 5. \end{cases}$$

The only x with $2 < x \leq 4.8$ and non-zero probability is $x = 3$. Then $P(2 < X \leq 4.8) = P(X = 3) = 1/10$. Analogously, $P(2 \leq X \leq 4.8) = P(X = 2) + P(X = 3) = 2/10$. \diamond

- 3.** Let F be the CDF for a random variable X . Then:

- (a) $P(X = x) = F(x) - F(x^-)$ where $F(x^-) = \lim_{y \uparrow x} F(y)$;
- (b) $P(x < X \leq y) = F(y) - F(x)$;
- (c) $P(X > x) = 1 - F(x)$;
- (d) If X is continuous, then

$$F(b) - F(a) = P(a < X < b) = P(a \leq X < b) = P(a < X \leq b) = P(a \leq X \leq b).$$

Proof. (a) Clearly, $\lim_{y \uparrow x} \{X \leq y\} = \{X < x\}$. Then (by the continuity of probabilities theorem)

$$F(x^-) = \lim_{y \uparrow x} P(X \leq y) = P(X < x) = P(X \leq x) - P(X = x) = F(x) - P(X = x).$$

- (b) Since $\{X \leq x\} \sqcup \{x < X \leq y\} = \{X \leq y\}$, then

$$P(x < X \leq y) = P(X \leq y) - P(X \leq x) = F(y) - F(x).$$

- (c) Since $\Omega = \{X > x\} \sqcup \{X \leq x\}$, then

$$1 = P(\Omega) = P(X > x) + P(X \leq x) = P(X > x) + F(x).$$

- (d) If X is continuous, $P(X = x) = 0$ for all x . Using this and expressing the intervals as disjoint unions (such as $\{a \leq X < b\} = \{X = a\} \sqcup \{a < X < b\}$) we obtain the desired result. \square

- 4.** Let X have pdf

$$f(x) = \begin{cases} 1/4 & \text{if } 0 < x < 1, \\ 3/8 & \text{if } 3 < x < 5, \\ 0 & \text{otherwise.} \end{cases}$$

- (a) Find the CDF of X .
(b) Let $Y = 1/X$. Find the pdf of Y .

Solution. (a) By definition,

$$F(x) = \int_{-\infty}^x f(t) dt = \begin{cases} 0 & \text{if } x \leq 0, \\ (1/4)x & \text{if } 0 < x < 1, \\ 1/4 & \text{if } 1 \leq x \leq 3, \\ 1/4 + (3/8)(x - 3) & \text{if } 3 < x < 5, \\ 1 & \text{if } x \geq 5. \end{cases}$$

- (b) Since $r = 1/x$ is monotone decreasing in $(0, \infty)$ it has an inverse $s = 1/y$ and

$$f_Y(y) = f_X(s(y)) \left| \frac{ds(y)}{dy} \right| = \begin{cases} \frac{3}{8y^2} & \text{if } 1/5 < y < 1/3, \\ \frac{1}{4y^2} & \text{if } y > 1, \\ 0 & \text{otherwise.} \end{cases}$$

◊

5. Let X and Y be discrete random variables. Show that X and Y are independent if and only if

$$f_{X,Y}(x, y) = f_X(x)f_Y(y) \quad (1)$$

for all x, y .

Proof. If X and Y are independent, eq. (1) is trivial. Conversely, suppose eq. (1). Then for any A, B ,

$$\begin{aligned} P(X \in A, Y \in B) &= \sum_{x \in A} \sum_{y \in B} f(x, y) = \sum_{x \in A} \sum_{y \in B} f(x)f(y) = \\ &\sum_{x \in A} \left(f(x) \sum_{y \in B} f(y) \right) = \left(\sum_{x \in A} f(x) \right) \left(\sum_{y \in B} f(y) \right) = P(X \in A)P(Y \in B). \end{aligned}$$

□

6. Let X have distribution F and density f . Let $A \subset \mathbb{R}$ and let $Y = \chi_A(X)$ where χ_A is the characteristic function of A . Find the CDF of Y .

Solution. By definition, Y has a Bernoulli distribution with success probability

$$P(Y = 1) = P(\chi_A(X) = 1) = P(X \in A) = \int_A f(x) dx.$$

Therefore,

$$F_Y(y) = \begin{cases} 0 & \text{if } y < 0, \\ 1 - \int_A f(x) dx & \text{if } 0 \leq y < 1, \\ 1 & \text{if } y \geq 1. \end{cases}$$

◊

7. Let X and Y be independent and suppose that each has Uniform(0, 1) distribution. Let $Z = \min\{X, Y\}$. Find the density of Z .

Solution. Following the hint,

$$P(Z > z) = P(\min\{X, Y\} > z) = P(X > z, Y > z) = P(X > z)P(Y > z) = (1 - F_X(z))(1 - F_Y(z))$$

where the third equality is because of independence. Furthermore,

$$P(Z \leq z) = 1 - (1 - F_X(z))(1 - F_Y(z))$$

Then, since the CDF of a Uniform(0, 1) distributed random variable is the identity in [0, 1],

$$F_Z(z) = 1 - (1 - z)^2 = 2z - z^2 \text{ for } 0 \leq z \leq 1$$

In particular,

$$f_Z(z) = 2 - 2z \text{ for } 0 \leq z \leq 1.$$

◇

8. Let X have CDF F . Find the CDF of $X^+ = \max\{0, X\}$.

Solution. We distinguish two cases:

$$P(X^+ \leq z) = \begin{cases} P(X \leq z) & \text{if } z \geq 0, \\ 0 & \text{otherwise.} \end{cases}$$

◇

10. Let X and Y be independent. Prove that $g(X)$ and $h(Y)$ are independent, where g and h are functions.

Proof. Simply use $\{g(X) \in A\} = \{X \in g^{-1}(A)\}$. □

11. (a) Suppose we toss a coin once and let p be the probability of heads. Let X and Y be the number of heads and tails. Prove that X and Y are dependent.

(b) Let $N \sim \text{Poisson}(\lambda)$ and suppose we toss a coin N times. Let X and Y be the number of heads and tails. Show that X and Y are independent.

Proof. (a) Note that

$$P(X = 1, Y = 1) = 0 < P(X = 1)P(Y = 1).$$

(b) Calculating directly,

$$P(X = x, Y = y) = P(X = x \mid N = x + y)P(N = x + y) = \binom{x+y}{x} p^x (1-p)^y \left(e^{-\lambda} \frac{\lambda^{x+y}}{(x+y)!} \right)$$

since $X \mid N \sim \text{Binomial}(N, p)$. This expression may be factored as a product of the form $g(x)h(y)$ (the $(x+y)!$'s will cancel each other) and therefore, X and Y are independent. □

12. Suppose the range of X and Y is a (possibly infinite) rectangle. If $f_{X,Y}(x, y) = g(x)h(y)$ for some functions g and h (not necessarily pdfs) then X and Y are independent.

Proof. Note that

$$f_X(x) = \int f_{X,Y}(x, y) dy = \int g(x)h(y) dy = g(x) \int h(y) dy.$$

Analogously,

$$f_Y(y) = h(y) \int g(x) dx.$$

Therefore,

$$f_X(x)f_Y(y) = \left(g(x) \int h(y) dy \right) \left(h(y) \int g(x) dx \right) = g(x)h(y)$$

where the last equality follows from $1 = \iint f(x, y) dxdy = \int g(x) dx \int h(y) dy$ (which is possible because the domain is a rectangle and because of Fubini). \square

13. Let $X \sim N(0, 1)$ and let $Y = e^X$. Find the pdf of Y .

Proof. Since $r = e^x$ is strictly increasing in \mathbb{R} , it has an inverse $s = \log y$ and

$$f_Y(y) = f_X(s(y)) \left| \frac{ds(y)}{dy} \right| = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{(\log y)^2}{2}\right) \left| \frac{1}{y} \right|.$$

\square

14. Let (X, Y) be uniformly distributed on the unit disk $D^2 = \{(x, y) : x^2 + y^2 \leq 1\}$. Let $R = \sqrt{X^2 + Y^2}$. Find the CDF and PDF of R .

Proof. Note that

$$A_r = \{(x, y) \in D^2 : \sqrt{x^2 + y^2} \leq r\} = \begin{cases} \emptyset & \text{if } r \leq 0, \\ B_r(0) & \text{if } 0 < r < 1, \\ D^2 & \text{if } r \geq 1 \end{cases}$$

Therefore,

$$\begin{aligned} F_R(r) &= \iint_{A_r} f(x, y) dxdy \\ &= \begin{cases} \iint_{\emptyset} 1/\pi dxdy & \text{if } r \leq 0, \\ \iint_{B_r(0)} 1/\pi dxdy & \text{if } 0 < r < 1, \\ \iint_{D^2} 1/\pi dxdy & \text{if } r \geq 1. \end{cases} \\ &= \begin{cases} 0 & \text{if } r \leq 0, \\ r^2 & \text{if } 0 < r < 1, \\ 1 & \text{if } r \geq 1. \end{cases} \end{aligned}$$

Furthermore,

$$f_R(r) = 2r \text{ for } 0 < r < 1.$$

\square

15. Let X have a continuous, strictly increasing CDF F . Let $Y = F(X)$. Find the density of Y . Now let $U \sim \text{Uniform}(0, 1)$ and let $X = F^{-1}(U)$. Show that $X \sim F$. (This result still holds without the continuous, strictly increasing hypothesis.)

Proof. Note that Y only takes values in $[0, 1]$. Then $F_Y(y) = 0$ if $y \leq 0$ and $F_Y(y) = 1$ if $y \geq 1$ and

$$F_Y(y) = P(Y \leq y) = P(F(X) \leq y) = P(X \leq F^{-1}(y)) = F(F^{-1}(y)) = y \text{ if } 0 < y < 1.$$

Therefore,

$$f(y) = 1 \text{ for } 0 < y < 1$$

and $Y \sim \text{Uniform}(0, 1)$. Conversely,

$$F_X(x) = P(X \leq x) = P(F^{-1}(U) \leq x) = P(U \leq F(x)) = F(x).$$

That is, $X \sim F$. \square

16. Let $X \sim \text{Poisson}(\lambda)$ and $Y \sim \text{Poisson}(\mu)$ and assume X and Y are independent. Show that the distribution of X given $X + Y = n$ is Binomial(n, π) where $\pi = \lambda/(\lambda + \mu)$.

Proof. Following the hint, for $x \in \mathbb{N}$,

$$\begin{aligned} P(X = x \mid X + Y = n) &= \frac{P(X = x, X + Y = n)}{P(X + Y = n)} \\ &= \frac{P(X = x, Y = n - x)}{P(X + Y = n)} \\ &= \frac{\left(e^{-\lambda} \frac{\lambda^x}{x!}\right) \left(e^{-\mu} \frac{\mu^{n-x}}{(n-x)!}\right)}{e^{-(\lambda+\mu)} \frac{(\lambda+\mu)^n}{n!}} \\ &= \frac{\left(\frac{\lambda^x}{x!}\right) \left(\frac{\mu^{n-x}}{(n-x)!}\right)}{\frac{(\lambda+\mu)^n}{n!}} \\ &= \binom{n}{x} \pi^x (1 - \pi)^{n-x}. \end{aligned}$$

Remark. The trick $\{X = x, X + Y = n\} = \{X = x, Y = n - x\}$ looks very useful. \square

17. Let

$$f_{X,Y}(x, y) = \begin{cases} c(x + y^2) & \text{if } 0 \leq x \leq 1 \text{ and } 0 \leq y \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

Find $P(X < 1/2 \mid Y = 1/2)$.

Solution. By definition,

$$f_Y(y) = \int f_{X,Y}(x, y) dx = c \int_0^1 (x + y^2) dx = c(1/2 + y^2).$$

Then

$$f(x \mid y) = \frac{f(x, y)}{f(y)} = \frac{x + y^2}{1/2 + y^2}.$$

Finally,

$$P(X < 1/2 \mid Y = 1/2) = \int_0^{1/2} f(x \mid 1/2) dx = \int_0^{1/2} \frac{x + 1/4}{1/2 + 1/4} dx = 1/3.$$

\diamond

19. Let X be continuous and $Y = r(X)$. If r is strictly increasing or decreasing, then r has an inverse $s = r^{-1}$ and

$$f_Y(y) = f_X(s(y)) \left| \frac{ds(y)}{dy} \right|.$$

Proof. Suppose r is strictly decreasing. Then

$$\begin{aligned} F_Y(y) &= P(Y \leq y) = P(r(X) \leq y) = \\ P(X \geq r^{-1}(y)) &= 1 - P(X < r^{-1}(y)) = 1 - F_X(r^{-1}(y)). \end{aligned}$$

Differentiating yields,

$$f_Y(y) = -f_X(s(y)) \frac{ds(y)}{dy} = f_X(s(y)) \left| \frac{ds(y)}{dy} \right|$$

where the last equality is because s is decreasing (since r is too). The other case is analogous. \square