

# All of Statistics - Chapter 2 Solutions

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**Acknowledgements:** Thanks to Ben S. for correcting some mistakes.

**1.**

By Lemma 2.15,  $\mathbb{P}(X = x) = F(x) - F(x-)$ . Since  $F$  is right-continuous,  $F(x) = F(x+)$ .

**2.**

By Lemma 2.15,

$$\mathbb{P}(2 < X \leq 4.8) = F(4.8) - F(2) = 1/10$$

and

$$\mathbb{P}(2 \leq X \leq 4.8) = \mathbb{P}(X = 2) + \mathbb{P}(2 < X \leq 4.8) = F(4.8) - F(2-) = 2/10.$$

**3.**

**1)**

Since  $F$  is monotone, we can write  $F(x-) = \lim_n F(x_n)$  where  $(x_n)$  is some strictly increasing sequence converging to  $x$ . Let  $A_n = \{X \leq x_n\}$  so that  $\{X < x\} = \cup_n A_n$ . By continuity of probability,  $\mathbb{P}(X < x) = \lim_k \mathbb{P}(A_n) = \lim_n F(x_n)$ .

**2)**

By additivity,  $\mathbb{P}(X \leq x) + \mathbb{P}(x < X \leq y) = \mathbb{P}(X \leq y)$ . The desired result follows by moving some terms around.

**3)**

Taking complements,  $\mathbb{P}(X > x) = 1 - \mathbb{P}(X \leq x) = 1 - F(x)$ .

**4)**

If  $X$  is continuous,  $\mathbb{P}(X = x) = 0$  for all  $x$  by Part 1. The desired result follows from combining this fact with the findings from Part 2.

**4.**

**a)**

We can express the CDF using indicator functions:

$$F_X(x) = \frac{x}{4} I_{[0,1)}(x) + \frac{1}{4} I_{[1,\infty)}(x) + \frac{3}{8}(x-3) I_{[3,5)}(x) + \frac{3}{4} I_{[5,\infty)}(x).$$

**b)**

Since  $Y = 1/X$  and  $F_X(0) = 0$ , it follows that  $F_Y(0) = 0$ . For  $y > 0$ ,

$$F_Y(y) = \mathbb{P}(X \geq 1/y) = 1 - \mathbb{P}(X < 1/y) = 1 - F_X(1/y).$$

**5.**

Suppose  $X$  and  $Y$  are independent. Then,

$$f_{X,Y}(x, y) = \mathbb{P}(X \in \{x\}, Y \in \{y\}) = \mathbb{P}(X \in \{x\})\mathbb{P}(Y \in \{Y\}) = f_X(x)f_Y(y).$$

To establish the converse, suppose that  $f_{X,Y} = f_X f_Y$ . For a subset  $A$  of the support of  $X$  and a subset  $B$  of the support of  $Y$ ,

$$\begin{aligned} \mathbb{P}(X \in A, Y \in B) &= \sum_{(x,y) \in A \times B} f_{X,Y}(x, y) = \sum_{x \in A} f_X(x) \sum_{y \in B} f_Y(y) \\ &= \mathbb{P}(X \in A)\mathbb{P}(Y \in B). \end{aligned}$$

**6.**

Note that

$$F_Y(y) = \begin{cases} 0 & \text{if } y < 0 \\ \mathbb{P}(X \notin A) & \text{if } 0 \leq y < 1 \\ 1 & \text{if } y \geq 1. \end{cases}$$

**7.**

Since

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

it follows that

$$F_Z(z) = 1 - (1 - F_X(z))(1 - F_Y(z)) = F_X(z) + F_Y(z) - F_X(z)F_Y(z).$$

When  $X$  and  $Y$  have the same distribution  $F$ ,  $F_Z(z) = 2F(z) - F(z)^2$  and hence  $f_Z(z) = 2f(z) - 2F(z)f(z)$ . In particular, when  $F$  is a uniform distribution on  $(0, 1)$ ,

$$f_Z(z) = 2(1-z)I_{(0,1)}(z).$$

**8.**

Let  $Y = X^+$ . First, note that  $F_Y(0-) = 0$  and  $F_Y(0) = F_X(0)$ . Moreover,  $F_Y(x) = F_X(x)$  for  $x > 0$ .

**9.**

For  $x > 0$ ,  $F_X(x) = \int_0^x \lambda e^{-\lambda t} dt = 1 - e^{-\lambda x}$ . Therefore,  $F^{-1}(q) = -\ln(1 - q)/\lambda$ .

**10.**

If  $X$  and  $Y$  are independent, then

$$\begin{aligned} \mathbb{P}(g(X) \in A, h(Y) \in B) &= \mathbb{P}(X \in g^{-1}(A), Y \in h^{-1}(B)) \\ &= \mathbb{P}(X \in g^{-1}(A))\mathbb{P}(Y \in h^{-1}(B)) = \mathbb{P}(g(X) \in A)\mathbb{P}(h(Y) \in B) \end{aligned}$$

under some lax conditions on  $g$  and  $h$  (Borel measurable).

**11.**

**a)**

The two variables are dependent because

$$\mathbb{P}(X = 1, Y = 0) = 0 \neq p(1 - p) = \mathbb{P}(X = 1)\mathbb{P}(Y = 0).$$

**b)**

The two variables are independent because

$$\mathbb{P}(X = i, Y = j) = \frac{\lambda^{i+j} e^{-\lambda}}{(i+j)!} \binom{i+j}{i} p^i (1-p)^j = e^{-\lambda} \frac{\lambda^i p^i}{i!} \frac{\lambda^j (1-p)^j}{j!}$$

is decomposable into the form  $g(i)h(j)$ .

**12.**

If  $X$  and  $Y$  admit a joint density satisfying  $f(x, y) = g(x)h(y)$ , then

$$\mathbb{P}(X \leq x, Y \leq y) = \int_{-\infty}^x \int_{-\infty}^y f(s, t) dt ds = \int_{-\infty}^x g(s) ds \int_{-\infty}^y h(t) dt.$$

The marginal distribution for  $X$  is  $\mathbb{P}(X \leq x) = c_h \int_{-\infty}^x g(s) ds$  where  $c_h = \int_{-\infty}^{\infty} h(t) dt$ . It follows that  $f_X = hc_h$ . We can similarly define  $c_g$  to find that  $f_Y = gc_g$ . Moreover,  $c_h c_g = 1$  and hence  $c_g = 1/c_h$ . It follows that  $f_{X,Y} = f_X f_Y$ , as desired.

**13.**

**a)**

Note that

$$F_Y(y) = \mathbb{P}(e^X \leq y) = \mathbb{P}(X \leq \ln y) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\ln y} \exp\left(-\frac{x^2}{2}\right) dy.$$

Taking derivatives,

$$f_Y(y) = \frac{1}{y\sqrt{2\pi}} \exp\left(-\frac{(\ln y)^2}{2}\right).$$

**b)**

TODO (Computer Experiment)

**14.**

Let  $0 < r < 1$ . Then,  $F_R(r) = \pi r^2 / \pi = r^2$  and hence  $f_R(r) = 2r$ .

**15.**

For  $0 \leq y \leq 1$ ,

$$F_Y(y) = \mathbb{P}(F(X) \leq y) = \mathbb{P}(X \leq F^{-1}(y)) = F(F^{-1}(y)) = y.$$

For all  $x$ ,

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

**16.**

Note that

$$\mathbb{P}(X = x \mid X + Y = n) = \frac{\mathbb{P}(X = x, Y = n - x)}{\mathbb{P}(X + Y = n)}.$$

Moreover,

$$\mathbb{P}(X = x, Y = n - x) = \frac{e^{-\lambda} \lambda^x}{x!} \frac{e^{-\mu} \mu^{n-x}}{(n-x)!}.$$

As per the hint,

$$\mathbb{P}(X + Y = n) = e^{-\lambda - \mu} \frac{(\lambda + \mu)^n}{n!}.$$

Letting  $\pi = \lambda/(\lambda + \mu)$ , combining these facts yields

$$\mathbb{P}(X = x \mid X + Y = n) = \binom{n}{x} \pi^x (1 - \pi)^{n-x}.$$

## 17.

First, note that

$$f_Y(1/2) = \int_0^1 f(x, 1/2) dx = c \int_0^1 \left(x + \frac{1}{4}\right) dx = \frac{3}{4}c.$$

Therefore,

$$f_{X|Y}(x \mid 1/2) = \frac{f_{X,Y}(x, 1/2)}{f_Y(1/2)} = \frac{4}{3} \left(x + \frac{1}{4}\right) I_{(0,1)}(x).$$

It follows that

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

## 18.

TODO (Computer Experiment)

## 19.

Let  $r$  be strictly increasing with differentiable inverse  $s$ . Let  $X$  be an (absolutely) continuous random variable. Then, for  $Y = r(X)$ ,

$$F_Y(y) = \mathbb{P}(r(X) \leq y) = \mathbb{P}(X \leq s(y)) = F_X(s(y))$$

and hence  $f_Y(y) = f_X(s(y))s'(y)$ . If  $r$  was instead strictly decreasing, then

$$F_Y(y) = \mathbb{P}(X \geq s(y)) = 1 - F_X(s(y))$$

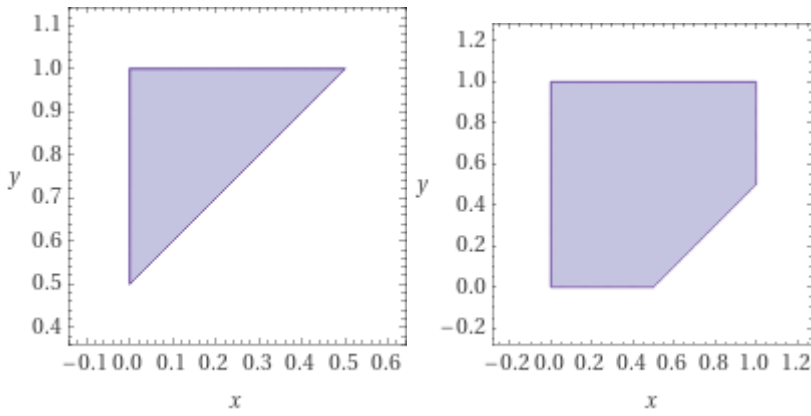
and hence  $f_Y(y) = -f_X(s(y))s'(y)$ . Since a strictly decreasing function has a strictly decreasing inverse, it follows that  $s'(y) < 0$  and hence we can summarize both cases by  $f_Y = (f_X \circ s)|s'|$ .

## 20.

Let  $W = X - Y$ . Then,  $F_W(-1) = 0$  and  $F_W(1) = 1$ . For  $-1 < w < 1$ ,  $F_W(w) = \mathbb{P}(Y \geq X - w)$ . The region

$$\{(x, y) : y \geq x - w, 0 \leq x, y \leq 1\}$$

is either a triangle or a right trapezoid depending on whether  $-1 < w < 0$  or  $0 < w < 1$ :



By covering these case separately, one can derive  $F_W(w) = (1 + w)^2/2$  and  $F_W(w) = -w^2/2 + w + 1/2$ , respectively. It follows that

$$f_W(w) = \begin{cases} 1 + w & \text{if } -1 < w < 0 \\ 1 - w & \text{if } 0 < w < 1 \\ 0 & \text{otherwise.} \end{cases}$$

Let  $V = X/Y$ . Then,  $F_V(0) = 0$ . For  $v > 0$ ,  $F_V(v) = \mathbb{P}(Y \geq X/v)$ . The region

$$\left\{ (x, y) : y \geq \frac{x}{v}, 0 \leq x, y \leq 1 \right\}$$

is either a triangle or a rectangle plus a right trapezoid depending on whether  $0 < v < 1$  or  $v > 1$ . By covering these cases separately, one can derive  $F_V(v) = 2/v$  and  $F_V(v) = 1/(2v) + (1 - 1/v)$ , respectively. It follows that

$$f_V(v) = \begin{cases} 1/2 & \text{if } 0 < v < 1 \\ 1/(2v^2) & \text{if } v > 1 \\ 0 & \text{otherwise.} \end{cases}$$

## 21.

Since

$$F_Y(y) = \mathbb{P}(\max\{X_1, \dots, X_n\} \leq y) = \mathbb{P}(X_1 \leq y)^n = (1 - e^{-\beta y})^n,$$

it follows that  $f_Y(y) = \beta n e^{-\beta y} (1 - e^{-\beta y})^{n-1}$ .