

**CASELLA-BERGER  
STATISTICAL INFERENCE SOLUTION:  
CHAPTER 2**

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## 1. PROBLEM 2.1

In each of the following find the pdf of  $Y$ . Show that the pdf integrates to 1.

- (a)  $Y = X^3$  and  $f_X(x) = 42x^5(1-x)$ ,  $0 < x < 1$
- (b)  $Y = 4X + 3$  and  $f_X(x) = 7e^{-7x}$ ,  $0 < x < \infty$
- (c)  $Y = X^2$  and  $f_X(x) = 30x^2(1-x)^2$ ,  $0 < x < 1$

*Solution.* We begin by noting all conditions of [BC01, Theorem 2.1.5 on page 51] are satisfied in each case. We leave it to the reader to verify the pdf integrates to 1.

- (a) Let  $g(x) = x^3$  for  $x \in (0, 1)$ , then  $g^{-1}(y) = y^{\frac{1}{3}}$  for  $y \in (0, 1)$ , and

$$\left| \frac{d}{dy} g^{-1}(y) \right| = \frac{1}{3y^{\frac{2}{3}}}.$$

Hence,

$$\begin{aligned} f_Y(y) &= f_X(g^{-1}(y)) \cdot \left| \frac{d}{dy} g^{-1}(y) \right| \\ &= \left[ 42y^{\frac{5}{3}} \left( 1 - y^{\frac{1}{3}} \right) \right] \cdot \frac{1}{3y^{\frac{2}{3}}} \\ &= 14 \left( y - y^{\frac{4}{3}} \right) \end{aligned}$$

on  $\mathcal{Y} = (0, 1)$ .

- (b) Let  $g(x) = 4x + 3$  for  $x \in (0, \infty)$ , then  $g^{-1}(y) = \frac{y-3}{4}$  for  $y \in (3, \infty)$ , and

$$\left| \frac{d}{dy} g^{-1}(y) \right| = \frac{1}{4}.$$

Hence,

$$\begin{aligned} f_Y(y) &= f_X(g^{-1}(y)) \cdot \left| \frac{d}{dy} g^{-1}(y) \right| \\ &= \frac{7}{4} e^{-\frac{7(y-3)}{4}} \end{aligned}$$

on  $\mathcal{Y} = (3, \infty)$ .

- (c) Let  $g(x) = x^2$  for  $x \in (0, 1)$ , then  $g^{-1}(y) = y^{\frac{1}{2}}$  for  $y \in (0, 1)$ , and

$$\left| \frac{d}{dy} g^{-1}(y) \right| = \frac{1}{2} y^{-\frac{1}{2}}.$$

Hence,

$$\begin{aligned} f_Y(y) &= f_X(g^{-1}(y)) \cdot \left| \frac{d}{dy} g^{-1}(y) \right| \\ &= 15y^{\frac{1}{2}} \left(1 - y^{\frac{1}{2}}\right)^2 \end{aligned}$$

□

## 2. PROBLEM 2.2

In each of the following find the pdf of  $Y$ .

- (a)  $Y = X^2$  and  $f_X(x) = 1$ ,  $0 < x < 1$   
 (b)  $Y = -\log(X)$  and  $X$  has pdf

$$f_X(x) = \frac{(n+m+1)!}{n!m!} x^n (1-x)^m, \quad 0 < x < 1, \quad m, n \text{ positive integers}$$

- (c)  $Y = e^X$  and  $X$  has pdf

$$f_X(x) = \frac{1}{\sigma^2} x e^{-\frac{(x/\sigma)^2}{2}}, \quad 0 < x < \infty, \quad \sigma^2 \text{ a positive constant}$$

*Solution.* We begin by noting all conditions of [BC01, Theorem 2.1.5 on page 51] are satisfied in each case.

- (a) Let  $g(x) = x^2$  for  $x \in (0, 1)$ , then  $g^{-1}(y) = y^{\frac{1}{2}}$  for  $y \in (0, 1)$ , and

$$\left| \frac{d}{dy} g^{-1}(y) \right| = \frac{1}{2} y^{-\frac{1}{2}}.$$

Hence,

$$\begin{aligned} f_Y(y) &= f_X(g^{-1}(y)) \cdot \left| \frac{d}{dy} g^{-1}(y) \right| \\ &= \frac{1}{2} y^{-\frac{1}{2}} \end{aligned}$$

on  $\mathcal{Y} = (0, 1)$ .

- (b) Let  $g(x) = -\log(x)$  for  $x \in (0, 1)$ , then  $g^{-1}(y) = e^{-y}$  for  $y \in (0, \infty)$ , and

$$\left| \frac{d}{dy} g^{-1}(y) \right| = e^{-y}.$$

Hence,

$$\begin{aligned} f_Y(y) &= f_X(g^{-1}(y)) \cdot \left| \frac{d}{dy} g^{-1}(y) \right| \\ &= \frac{(n+m+1)!}{n!m!} e^{-ny} (1 - e^{-y})^m \end{aligned}$$

on  $\mathcal{Y} = (0, \infty)$ .

- (c) Let  $g(x) = e^x$  for  $x \in (0, \infty)$ , then  $g^{-1}(y) = \log(y)$  for  $y \in (1, \infty)$ , and

$$\left| \frac{d}{dy} g^{-1}(y) \right| = \frac{1}{y}.$$

Hence,

$$\begin{aligned} f_Y(y) &= f_X(g^{-1}(y)) \cdot \left| \frac{d}{dy} g^{-1}(y) \right| \\ &= \frac{\log(y)}{y\sigma^2} e^{-\frac{(\log(y)/\sigma)^2}{2}} \end{aligned}$$

on  $\mathcal{Y} = (0, \infty)$ .

□

## 3. PROBLEM 2.3

Suppose  $X$  has the geometric pmf  $f_X(x) = \frac{1}{3} \left(\frac{2}{3}\right)^x$ ,  $x = 0, 1, 2, \dots$ . Determine the probability distribution of  $Y = X/(X+1)$ . Note that here both  $X$  and  $Y$  are discrete random variables. To specify the probability distribution of  $Y$ , specify its pmf.

*Solution.*

$$\begin{aligned}
 f_Y(y) &= P(Y = y) \\
 &= P\left(\frac{X}{X+1} = y\right) \\
 &= P\left(X = \frac{y}{1-y}\right) \\
 &= f_X\left(\frac{y}{1-y}\right) \\
 &= \frac{1}{3} \left(\frac{2}{3}\right)^{\frac{y}{1-y}}
 \end{aligned}$$

$$\text{on } \mathcal{Y} = \left\{ \frac{x}{x+1} \mid x = \frac{1}{3} \left(\frac{2}{3}\right)^k \text{ for some } k \in \mathbb{N} \cup \{0\} \right\}$$

□

## 4. PROBLEM 2.4

Let  $\lambda$  be a fixed positive constant, and define the function  $f(x)$  by  $f(x) = \frac{1}{2}\lambda e^{-\lambda x}$  if  $x \geq 0$  and  $f(x) = \frac{1}{2}\lambda e^{\lambda x}$  if  $x < 0$ .

- (a) Verify that  $f(x)$  is a pdf.
- (b) If  $X$  is a random variable with pdf given by  $f(x)$ , find  $P(X < t)$  for all  $t$ . Evaluate all integrals.
- (c) Find  $P(|X| < t)$  for all  $t$ . Evaluate all integrals.

*Solution.*

- (a) Check the conditions listed on [BC01, Theorem 1.6.5 on page 36] for  $f(x)$ .

(b)

$$\begin{aligned}
 P(X < t) &= \int_{-\infty}^t f(x) \, dx \\
 &= \begin{cases} \int_{-\infty}^t \frac{1}{2}\lambda e^{\lambda x} \, dx & \text{if } t < 0, \\ \int_{-\infty}^0 \frac{1}{2}\lambda e^{\lambda x} \, dx + \int_0^t \frac{1}{2}\lambda e^{-\lambda x} \, dx & \text{if else,} \end{cases} \\
 &= \begin{cases} \frac{e^{\lambda t}}{2} & \text{if } t < 0, \\ 1 - \frac{1}{2}e^{-\lambda t} & \text{if else.} \end{cases}
 \end{aligned}$$

(c)

$$\begin{aligned}
 P(|X| < t) &= P(-t < X < t) \\
 &= P(X < t) - P(X < -t) \\
 &= \left(1 - \frac{1}{2}e^{-\lambda t}\right) - \frac{e^{-\lambda t}}{2} \quad \text{(part (b))} \\
 &= 1 - e^{-\lambda t}
 \end{aligned}$$

□

## 5. PROBLEM 2.5

Use [BC01, Theorem 2.1.8 on page 53] to find the pdf of  $Y$  in [BC01, Example 2.1.2 on page 49]. Show that the same answer is obtained by differentiating the cdf given in [BC01, Equation 2.1.6 on page 49].

*Solution.* Partition the interval  $(0, 2\pi)$  into  $\{A_i\}_{i=0}^4$ , with

$$A_i = \begin{cases} \{0\} & \text{if } i = 0, \\ \left(\frac{(i-1)\pi}{2}, \frac{i\pi}{2}\right) & \text{if } i > 0. \end{cases}$$

For each  $i$ , write  $g_i(x) = \sin^2(x)$  on  $A_i$ . Then

$$g_1^{-1}(y) = \arcsin(\sqrt{y})$$

$$g_2^{-1}(y) = \pi - \arcsin(\sqrt{y})$$

$$g_3^{-1}(y) = \pi + \arcsin(\sqrt{y})$$

$$g_4^{-1}(y) = 2\pi - \arcsin(\sqrt{y})$$

Therefore,

$$\begin{aligned} f_Y(y) &= \sum_{i=1}^4 f_X(g_i^{-1}(y)) \cdot \left| \frac{d}{dy} g_i^{-1}(y) \right| \\ &= 4 \left( \frac{1}{2\pi} \right) \left[ \frac{1}{2\sqrt{y-y^2}} \right] \\ &= \frac{1}{\pi\sqrt{y-y^2}} \end{aligned}$$

on  $\mathcal{Y} = (0, 1)$ . □



## 6. PROBLEM 2.6

In each of the following find the pdf of  $Y$  and show that the pdf integrates to 1.

- (a)  $f_X(x) = \frac{1}{2}e^{-|x|}$ ,  $-\infty < x < \infty$ ;  $Y = |X|^3$   
 (b)  $f_X(x) = \frac{3}{8}(x+1)^2$ ,  $-1 < x < 1$ ;  $Y = 1 - X^2$   
 (c)  $f_X(x) = \frac{3}{8}(x+1)^2$ ,  $-1 < x < 1$ ;  $Y = 1 - X^2$  if  $X \leq 0$  and  $Y = 1 - X$  if  $X > 0$

*Solution.* We note that [BC01, Theorem 2.1.8 on page 53] applies to all cases, and let readers to verify the pdf integrates to 1.

- (a) Partition  $(-\infty, \infty)$  into

$$\begin{aligned} A_0 &= \{0\} \\ A_1 &= (-\infty, 0) \\ A_2 &= (0, \infty) \end{aligned}$$

and define

$$g_i(x) = \begin{cases} x^3 & \text{if } i \text{ even,} \\ -x^3 & \text{if } i \text{ odd} \end{cases}$$

on  $A_i$ . Then

$$\begin{aligned} f_Y(y) &= \sum_{i=1}^2 f_X(g_i^{-1}(y)) \cdot \left| \frac{d}{dy} g_i^{-1}(y) \right| \\ &= \frac{1}{3} y^{-\frac{2}{3}} e^{-y^{1/3}} \end{aligned}$$

on  $\mathcal{Y} = (0, \infty)$ .

- (b) Partition  $(-1, 1)$  into

$$\begin{aligned} A_0 &= \{0\} \\ A_1 &= (-1, 0) \\ A_2 &= (0, 1) \end{aligned}$$

and define

$$g_i(x) = 1 - x^2$$

on  $A_i$ . Then

$$\begin{aligned} f_Y(y) &= \sum_{i=1}^2 f_X(g_i^{-1}(y)) \cdot \left| \frac{d}{dy} g_i^{-1}(y) \right| \\ &= \frac{3}{8} \left( \frac{1}{\sqrt{1-y}} + \sqrt{1-y} \right) \end{aligned}$$

on  $\mathcal{Y} = (0, 1)$ .

(c) Partition  $(-1, 1)$  just as in part (b), and define

$$g_i(x) = \begin{cases} 1 - x^2 & \text{on } A_1, \\ 1 - x & \text{on } A_2. \end{cases}$$

Then

$$\begin{aligned} f_Y(y) &= \sum_{i=1}^2 f_X(g_i^{-1}(y)) \cdot \left| \frac{d}{dy} g_i^{-1}(y) \right| \\ &= \frac{3}{16} \frac{1}{\sqrt{1-y}} \left(1 - \sqrt{1-y}\right)^2 + \frac{3}{8} (2-y)^2 \end{aligned}$$

on  $\mathcal{Y} = (0, 1)$ .

□

## 7. PROBLEM 2.7

Let  $X$  have pdf  $f_X(x) = \frac{2}{9}(x+1)$ ,  $-1 \leq x \leq 2$ .

- (a) Find the pdf of  $Y = X^2$ . Note that [BC01, Theorem 2.1.8 on page 53] is not directly applicable in this problem.
- (b) Show that [BC01, Theorem 2.1.8 on page 53] remains valid if the sets  $A_0, A_1, \dots, A_k$  contain  $\mathcal{X}$ , and apply the extension to solve part (a) using  $A_0 = \emptyset$ ,  $A_1 = (-2, 0)$ , and  $A_2 = (0, 2)$ .

*Solution.*

(a)

$$\begin{aligned}
 P(Y \leq y) &= P(X^2 \leq y) \\
 &= \begin{cases} P(-\sqrt{y} \leq X \leq \sqrt{y}) & \text{if } y < 1, \\ P(-1 \leq X \leq \sqrt{y}) & \text{if } 1 \leq y \leq 4. \end{cases} \\
 &= \begin{cases} \int_{-\sqrt{y}}^{\sqrt{y}} f_X(x) dx & \text{if } y < 1, \\ \int_{-1}^{\sqrt{y}} f_X(x) dx & \text{if } 1 \leq y \leq 4. \end{cases} \\
 &= \begin{cases} \frac{4\sqrt{y}}{9} & \text{if } y < 1, \\ \frac{1}{9}(1 + \sqrt{y})^2 & \text{if } 1 \leq y \leq 4. \end{cases}
 \end{aligned}$$

on  $\mathcal{Y} = (0, 4)$ .

(b) C.f. Problem 2.6.

□

## 8. PROBLEM 2.8

In each of the following show that the given function is a cdf and find  $F_X^{-1}(y)$ .

(a)

$$F_X(x) = \begin{cases} 0 & \text{if } x < 0 \\ 1 - e^{-x} & \text{if } x \geq 0 \end{cases}$$

(b)

$$F_X(x) = \begin{cases} e^x/2 & \text{if } x < 0 \\ 1/2 & \text{if } 0 \leq x < 1 \\ 1 - (e^{1-x}/2) & \text{if } 1 \leq x \end{cases}$$

(c)

$$F_X(x) = \begin{cases} e^x/4 & \text{if } x < 0 \\ 1 - (e^{-x}/4) & \text{if } x \geq 0 \end{cases}$$

*Solution.* To show a function is cdf, we verify the conditions in [BC01, Theorem 1.5.3 on page 31], which are routine computations.

(a)

$$F_X^{-1}(y) = -\log(1 - y)$$

(b)

$$F_X^{-1}(y) = \begin{cases} \log(2y) & \text{if } 0 \leq y \leq \frac{1}{2} \\ 1 - \log(2(1 - y)) & \text{if } \frac{1}{2} \leq y \leq 1 \end{cases}$$

(c)

$$F_X^{-1}(y) = \begin{cases} \log(4y) & \text{if } 0 \leq y \leq \frac{1}{4} \\ -\log(4(1 - y)) & \text{if } \frac{1}{4} \leq y \leq 1 \end{cases}$$

□

## 9. PROBLEM 2.9

If the random variable  $X$  has pdf

$$f(x) = \begin{cases} \frac{x-1}{2} & 1 < x < 3, \\ 0 & \text{otherwise,} \end{cases}$$

find a monotone function  $u(x)$  such that the random variable  $Y = u(X)$  has uniform(0,1) distribution.

*Solution.* This is a direct application of [BC01, Theorem 2.1.10 on page 54]. The cdf is given by

$$\begin{aligned} F_X(x) &= \begin{cases} 0 & \text{if } x \leq 1 \\ \int_1^x f(t) dt & \text{if } 1 < x < 3 \\ 1 & \text{if else} \end{cases} \\ &= \begin{cases} 0 & \text{if } x \leq 1 \\ \frac{(x-1)^2}{4} & \text{if } 1 < x < 3 \\ 1 & \text{if else} \end{cases} \end{aligned}$$

which is clearly monotone. So  $u(x) = F_X(x)$ . □

## 10. PROBLEM 2.11

Let  $X$  have the standard normal pdf,  $f_X(x) = (1/\sqrt{2\pi})e^{-x^2/2}$ .

- (a) Find  $EX^2$  directly, and then by using the pdf of  $Y = X^2$  from [BC01, Example 2.1.7 on page 52] and calculating  $EY$ .  
 (b) Find the pdf of  $Y = |X|$ , and find its mean and variance.

*Solution.*

- (a) First we have

$$\begin{aligned} EX^2 &= \int_{-\infty}^{\infty} x^2 f_X(x) dx && ([\text{BC01, Definition 2.2.1 on page 55}]) \\ &= \int_{-\infty}^{\infty} \frac{x^2}{\sqrt{2\pi}} e^{-x^2/2} dx \\ &= \frac{1}{\sqrt{2\pi}} \left[ x e^{-x^2/2} \Big|_{x=-\infty}^{x=\infty} - \int_{-\infty}^{\infty} e^{-x^2/2} dx \right] \\ &= 1. \end{aligned}$$

Secondly, by [BC01, Example 2.1.7 on page 52], the pdf of  $Y$  is given by

$$\begin{aligned} f_Y(y) &= \frac{1}{2\sqrt{y}} [f_X(\sqrt{y}) + f_X(-\sqrt{y})] \\ &= \frac{1}{\sqrt{2\pi y}} e^{-y/2}. \end{aligned}$$

Therefore,

$$\begin{aligned} EY &= \int_0^{\infty} y f_Y(y) dy \\ &= \int_0^{\infty} \sqrt{\frac{y}{2\pi}} e^{-y/2} dy \\ &= 1 \end{aligned}$$

- (b) Using [BC01, Theorem 2.1.8 on page 53],  $Y$  has pdf

$$\begin{aligned} f_Y(y) &= f_X(y) + f_X(-y) \\ &= \sqrt{\frac{2}{\pi}} e^{-y^2/2} \end{aligned}$$

Therefore,

$$EY = \int_0^\infty y f_Y(y) dy = \sqrt{\frac{2}{\pi}}$$
$$\text{Var}(Y) = EY^2 - (EY)^2 = 1 - \frac{2}{\pi}$$

□

## 11. PROBLEM 2.12

See [BC01, page 77] for the problem statement.

*Solution.* We know

$$y = \underbrace{d \tan(x)}_{g(x)}$$

for  $x \in (0, \pi/2)$ , and

$$\begin{aligned} \frac{dg^{-1}}{dy} &= \frac{d}{dy} \arctan\left(\frac{y}{d}\right) \\ &= \frac{d}{d^2 + y^2}. \end{aligned}$$

Therefore, [BC01, Theorem 2.1.5 on page 51] gives

$$\begin{aligned} f_Y(y) &= f_X(g^{-1}(y)) \left| \frac{dg^{-1}}{dy} \right| \\ &= \frac{1}{\frac{\pi}{2} - 0} \cdot \frac{d}{d^2 + y^2} \\ &= \frac{2d}{\pi(d^2 + y^2)} \end{aligned}$$

on  $\mathcal{Y} = (0, \infty)$ , which is the Cauchy distribution. In particular,  $EY = \infty$ . □



## 12. PROBLEM 2.13

Consider a sequence of independent coin flips, each of which has probability  $p$  of being heads. Define a random variable  $X$  as the length of the run (of either heads or tails) started by the first trail. (For example,  $X = 3$  if either TTTH or HHHT is observed.) Find the distribution of  $X$ , and find  $EX$ .

*Solution.*  $X$  has pmf

$$P(X = k) = (1 - p)^k p + p^k (1 - p).$$

Therefore,

$$\begin{aligned} EX &= \sum_{k=1}^{\infty} k [(1 - p)^k p + p^k (1 - p)] \\ &= (1 - p)p \left[ \sum_{k=1}^{\infty} k (1 - p)^{k-1} + \sum_{k=1}^{\infty} k p^{k-1} \right] \\ &= (1 - p)p \left( \frac{1}{p^2} + \frac{1}{(1 - p)^2} \right) \end{aligned}$$

□

## 13. PROBLEM 2.14

- (a) Let  $X$  be a continuous, nonnegative random variable [ $f(x) = 0$  for  $x < 0$ ]. Show that

$$EX = \int_0^\infty [1 - F_X(x)] dx,$$

where  $F_X(x)$  is the cdf of  $X$ .

- (b) Let  $X$  be a discrete random variable whose range is the nonnegative integers. Show that

$$EX = \sum_{k=0}^{\infty} (1 - F_X(k)),$$

where  $F_X(k) = P(X \leq k)$ . Compare this with part (a).

*Solution.*

(a)

$$\begin{aligned} \int_0^\infty [1 - F_X(x)] dx &= \int_0^\infty P(X > x) dx \\ &= \int_0^\infty \int_x^\infty f_X(y) dy dx \\ &= \int_0^\infty \int_0^y f_X(y) dx dy \\ &= \int_0^\infty y f_X(y) dy \\ &= EX \end{aligned}$$

(b)

$$\begin{aligned} EX &= \sum_{k=0}^{\infty} kP(X = k) \\ &= \sum_{k=1}^{\infty} P(X = k) + \sum_{k=2}^{\infty} P(X = k) + \sum_{k=3}^{\infty} P(X = k) + \cdots \\ &= P(X > 0) + P(X > 1) + P(X > 2) + \cdots \\ &= \sum_{k=0}^{\infty} 1 - F_X(k) \end{aligned}$$

□

## 14. PROBLEM 2.18

Show that if  $X$  is a continuous random variable, then

$$\min_a E |X - a| = E |X - m|,$$

where  $m$  is the median of  $X$ .

*Solution.* The expected value of  $|X - a|$  is given by

$$\begin{aligned} E |X - a| &= \int_{-\infty}^{\infty} |x - a| f_X(x) dx \\ &= \int_a^{\infty} (x - a) f_X(x) dx - \int_{-\infty}^a (x - a) f_X(x) dx \end{aligned}$$

Differentiate with respect to  $a$  we have

$$\begin{aligned} \frac{d}{da} E |X - a| &= \frac{d}{da} \left[ \int_a^{\infty} (x - a) f_X(x) dx \right] - \frac{d}{da} \left[ \int_{-\infty}^a (x - a) f_X(x) dx \right] \\ &= \int_a^{\infty} \frac{\partial}{\partial a} [(x - a) f_X(x)] dx - \int_{-\infty}^a \frac{\partial}{\partial a} [(x - a) f_X(x)] dx \\ &= \int_a^{\infty} f_X(x) dx - \int_a^{\infty} f_X(x) dx \\ &= P(X \leq a) - P(X > a). \end{aligned}$$

In particular,

$$1 - 2P(X > a) = \frac{d}{da} E |X - a| = 1 - 2P(X \leq a).$$

Therefore, the solution to

$$\frac{d}{da} E |X - a| = 0$$

is the median  $m$ . Moreover,  $m$  is a minima because

$$\left. \frac{d^2}{da^2} \right|_{a=m} E |X - a| = 2f_X(m) > 0.$$

□

## 15. PROBLEM 2.19

Prove that

$$\frac{d}{da}E(X - a)^2 = 0 \iff EX = a$$

by differentiating the integral. Verify, using calculus, that  $a = EX$  is indeed a minimum. List the assumptions about  $F_X$  and  $f_X$  are needed.

*Solution.* We have

$$\begin{aligned} \frac{d}{da}E(X - a)^2 &= \frac{d}{da} \int_{-\infty}^{\infty} (x - a)^2 f_X(x) dx \\ &= \int_{-\infty}^{\infty} \frac{\partial}{\partial a} [(x - a)^2 f_X(x)] dx \\ &= -2 \int_{-\infty}^{\infty} (x - a) f_X(x) dx \\ &= -2E(X - a) \end{aligned}$$

Therefore,

$$\begin{aligned} \frac{d}{da}E(X - a)^2 = 0 &\iff -2E(X - a) = 0 \\ &\iff E(X - a) = 0 \\ &\iff EX = a. \end{aligned}$$

To verify  $a = EX$  is minimum, we compute the second derivative

$$\frac{d^2}{da^2}E(X - a)^2 = 2 > 0.$$

□

## 16. PROBOELM 2.21

Prove the “two-way” rule for expectations, [BC01, Equation (2.2.5) on page 58], which says  $Eg(X) = EY$  where  $Y = g(X)$ . Assume that  $g(x)$  is a monotone function.

*Solution.*

$$\begin{aligned} Eg(X) &= \int_{\mathbb{R}} g(x) f_X(x) \, dx \\ &= \int_{\mathbb{R}} y f_X(g^{-1}(y)) \cdot \frac{dg^{-1}}{dy} \, dy \\ &= \int_{\mathbb{R}} y f_Y(y) \, dy \\ &= EY \end{aligned}$$

□

## 17. PROBLEM 2.22

Let  $X$  have the pdf

$$f(x) = \frac{4}{\beta^3 \sqrt{\pi}} x^2 e^{-x^2/\beta^2}, \quad 0 < x < \infty, \quad \beta > 0.$$

- (a) Verify that  $f(x)$  is a pdf.
- (b) Find  $EX$  and  $\text{Var}(X)$ .

*Solution.*

- (a) [BC01, Theorem 1.6.5 on page 36].
- (b)

$$\begin{aligned} EX &= \int_0^\infty x f(x) \, dx \\ &= \int_0^\infty \frac{4}{\beta^3 \sqrt{\pi}} x^3 e^{-x^2/\beta^2} \, dx \\ &= \frac{4}{\beta^3 \sqrt{\pi}} \int_0^\infty x^3 e^{-x^2/\beta^2} \, dx \\ &= \left( \frac{4}{\beta^3 \sqrt{\pi}} \right) \left( -\frac{\beta^2}{2} \right) \left( - \int_0^\infty 2x e^{-x^2/\beta^2} \, dx \right) \\ &= \left( \frac{4}{\beta^3 \sqrt{\pi}} \right) \left( \frac{\beta^4}{2} \right) \\ &= \frac{2\beta}{\sqrt{\pi}} \end{aligned}$$

and similarly,

$$\begin{aligned} EX^2 &= \frac{3\beta^2}{2}, \\ \text{Var}(X) &= EX^2 - (EX)^2 \\ &= \beta^2 \left[ \frac{3}{2} - \frac{4}{\pi} \right] \end{aligned}$$

□

## 18. PROBLEM 2.23

Let  $X$  have the pdf

$$f(x) = \frac{1}{2}(1+x), \quad -1 < x < 1.$$

- (a) Find the pdf of  $Y = X^2$ .  
 (b) Find  $EY$  and  $\text{Var}(Y)$ .

*Solution.*

- (a) Define  $g_i(x) = x^2$  on  $A_1 = (-1, 0)$  and  $A_2 = (0, 1)$ . Then

$$\begin{aligned} f_Y(y) &= [f(-\sqrt{y}) + f(\sqrt{y})] \cdot \frac{1}{2\sqrt{y}} \\ &= \frac{1}{2\sqrt{y}} \end{aligned}$$

on  $\mathcal{Y} = (0, 1)$ .

- (b) We have

$$\begin{aligned} \int_0^1 y^n f_Y(y) \, dy &= \frac{1}{2} \int_0^1 y^{n-1/2} \, dy \\ &= \frac{1}{2n+1}. \end{aligned}$$

This gives

$$\begin{aligned} EY &= \frac{1}{3} \\ EY^2 &= \frac{1}{5} \\ \text{Var}(Y) &= \frac{4}{45} \end{aligned}$$

□

## 19. PROBLEM 2.26

Let  $f(x)$  be a pdf and let  $a$  be a number such that, for all  $\varepsilon > 0$ ,  $f(a + \varepsilon) = f(a - \varepsilon)$ . Such a pdf is said to be symmetric about the point  $a$ .

- (a) Give three examples of symmetric pdfs.
- (b) Show that if  $X \sim f(x)$ , symmetric, then the median of  $X$  (see Exercise 2.17) is the number  $a$ .
- (c) Show that if  $X \sim f(x)$ , symmetric and  $EX$  exists, then  $EX = a$ .
- (d) Show that  $f(x) = e^{-x}$ ,  $x \geq 0$ , is not a symmetric pdf.
- (e) Show that for the pdf in part (d), the median is less than the mean.

*Solution.*

- (a) Cauchy, Normal, Uniform.
- (b) By change of variable, we may assume  $a = 0$ . The statement thus becomes: the median of an even pdf is 0, which is obvious because

$$\begin{aligned}
 1 &= \int_{-\infty}^{\infty} f(x) \, dx \\
 &= \int_{-\infty}^0 f(x) \, dx + \int_0^{\infty} f(x) \, dx \\
 &= 2 \int_{-\infty}^0 f(x) \, dx \\
 &= 2P(X \leq 0)
 \end{aligned}$$

- (c) Following the same logic in part (b), the statement becomes: the expected value of an even pdf  $f(x)$  is 0.

This is true because the function  $xf(x)$  is odd, hence

$$\begin{aligned}
 EX &= \int_{-\infty}^{\infty} xf(x) \, dx \\
 &= 0
 \end{aligned}$$

- (d) If  $f(x) = e^{-x}$  were symmetric, then it would be symmetric at  $x = EX$  for  $X \sim f(x)$  by part (c). In particular,

$$\begin{aligned}
 EX &= \int_0^{\infty} xe^{-x} \, dx \\
 &= 1.
 \end{aligned}$$

However, a direct computation shows 1 is not the median of  $X$ , contradicting part (b). Therefore it is not symmetric.

- (e) As computed in part (d), the mean is 1. The claim follows from the computation



$$\begin{aligned}\int_0^{\text{mean}} f(x) \, dx &= \int_0^1 e^{-x} \, dx \\ &= \frac{e-1}{e} \\ &> \frac{1}{2}\end{aligned}$$

□

## 20. EXERCISE 2.32

We compute

$$\begin{aligned}\left.\frac{d}{dt}\right|_{t=0} S(t) &= \left.\frac{d}{dt}\right|_{t=0} \log(M_X(t)) \\ &= \frac{\dot{M}_X(0)}{M_X(0)} \\ &= EX,\end{aligned}$$

and

$$\begin{aligned}\left.\frac{d^2}{dt^2}\right|_{t=0} S(t) &= \frac{\ddot{M}_X(0) M_X(0) - \dot{M}_X^2(0)}{M_X^2(0)} \\ &= EX^2 - (EX)^2 \\ &= \text{Var}(X).\end{aligned}$$

## 21. EXERCISE 2.33

(a) The mgf is

$$\begin{aligned}
 M_X(t) &= \sum_{x=0}^{\infty} e^{tx} \cdot \frac{e^{-\lambda} \lambda^x}{x!} \\
 &= e^{-\lambda} \sum_{x=0}^{\infty} \frac{(\lambda e^t)^x}{x!} \\
 &= e^{-\lambda} \cdot e^{\lambda e^t} \\
 &= e^{\lambda(e^t - 1)}.
 \end{aligned}$$

The moments are

$$\begin{aligned}
 EX &= \left. \frac{d}{dt} M_X(t) \right|_{t=0} \\
 &= \left. \lambda e^t \cdot e^{\lambda(e^t - 1)} \right|_{t=0} \\
 &= \lambda,
 \end{aligned}$$

$$\begin{aligned}
 EX^2 &= \left. \frac{d^2}{dt^2} M_X(t) \right|_{t=0} \\
 &= \left. \lambda (1 + e^t \lambda) \cdot e^{\lambda(e^t - 1)} \right|_{t=0} \\
 &= \lambda^2 + \lambda.
 \end{aligned}$$

Therefore,

$$\text{Var}(X) = \lambda.$$

(b) The mgf is

$$\begin{aligned}
 M_X(t) &= \sum_{x=0}^{\infty} e^{tx} p(1-p)^x \\
 &= \sum_{x=0}^{\infty} p [e^t(1-p)]^x \\
 &= \frac{p}{1 - e^t(1-p)}
 \end{aligned}$$

The moments are

$$EX = \frac{1-p}{p},$$

$$EX^2 = \frac{(2-p)(1-p)}{p^2}$$

Therefore,

$$\text{Var}(X) = \frac{1-p}{p^2}.$$

(c) The mgf is

$$\begin{aligned} M_X(t) &= \int_{-\infty}^{\infty} e^{tx} \cdot \frac{e^{-(x-\mu)^2/(2\sigma^2)}}{\sqrt{2\pi}\sigma} dx \\ &= \int_{-\infty}^{\infty} \frac{e^{-(x^2-2x\mu+\mu^2-2\sigma^2tx)/(2\sigma^2)}}{\sqrt{2\pi}\sigma} dx \\ &= e^{\frac{2\mu\sigma^2t+\sigma^4t^2}{2\sigma^2}} \cdot \int_{-\infty}^{\infty} \frac{e^{-\frac{[x-(\mu+\sigma^2t)]^2}{2\sigma^2}}}{\sqrt{2\pi}\sigma} dx \\ &= e^{\mu t + \frac{\sigma^2 t^2}{2}}. \end{aligned}$$

The moments are

$$\begin{aligned} EX &= \left. \frac{d}{dt} \right|_{t=0} M_X(t) \\ &= \mu, \end{aligned}$$

$$\begin{aligned} EX^2 &= \left. \frac{d^2}{dt^2} \right|_{t=0} M_X(t) \\ &= \sigma^2 + \mu^2. \end{aligned}$$

Therefore,

$$\begin{aligned} \text{Var}(X) &= EX^2 - (EX)^2 \\ &= \sigma^2 \end{aligned}$$

## REFERENCES

- [BC01] Roger Berger and George Casella. *Statistical Inference*. 2nd edition. Florence, AL: Duxbury Press, June 2001.