

## Chapter 5. Distribution of Functions of Random Variables

### Three Basic Methods

- cdf method
- pdf method (i.e., transformation-of-variables method)
- mgf method

### CDF Method

Usually applicable when  $Y = T(X)$  is 1-dimensional continuous random variable. Also, this method works well to find the distribution of  $\max(X_1, \dots, X_n)$  or  $\min(X_1, \dots, X_n)$ .

1. Rewrite  $Y$  in terms of  $X$ :  $(Y \leq y) = \{T(X) \leq y\} = (X \in \mathcal{A}_y)$
2. Find the cdf of  $Y$  using the distribution of  $X$ :  $F_Y(y) = P(Y \leq y) = P(X \in \mathcal{A}_y)$
3. Find the pdf of  $Y$  from the cdf:  $f_Y(y) = \frac{\partial}{\partial y} F_Y(y)$

**Ex 1.**  $X \sim \text{uniform}(0, 1)$ . Find the distribution of  $Y = -\ln(1 - X)$ .

*Answer:* First, note that  $y > 0$ .

$$F(y) = P(Y \leq y) = P\{-\ln(1 - X) \leq y\} = P(X \leq 1 - e^{-y}) = \begin{cases} 1 - e^{-y}, & y > 0 \\ 0, & \text{otherwise} \end{cases}$$

$$f(y) = \frac{\partial}{\partial y} F(y) = \begin{cases} e^{-y}, & y > 0 \\ 0, & \text{otherwise} \end{cases}$$

i.e.,  $Y \sim \text{exponential}$  with  $(\lambda = 1)$ .

**Ex 2.** Let  $X_1, X_2, \dots, X_n$  be iid continuous random variables with pdf  $f(\cdot)$ . Find

- (a) distribution of  $Y = \max(X_1, \dots, X_n)$
- (b) distribution of  $Z = \min(X_1, \dots, X_n)$

*Answer:*

$$(a) F_Y(y) = P\{\max(X_1, \dots, X_n) \leq y\} = P(X_1 \leq y, \dots, X_n \leq y) = \prod_{i=1}^n P(X_i \leq y) = \{F_X(y)\}^n$$

$$f_Y(y) = \frac{\partial}{\partial y} F_Y(y) = n \{F_X(y)\}^{n-1} f_X(y)$$

$$\begin{aligned}
\text{(b) } F_Z(z) &= P\{\min(X_1, \dots, X_n) \leq z\} = 1 - P(X_1 > z, \dots, X_n > z) = 1 - \prod_{i=1}^n P(X_i > z) \\
&= 1 - \{1 - F_X(z)\}^n \\
f_Z(z) &= \frac{\partial}{\partial z} F_Z(z) = n \{1 - F_X(z)\}^{n-1} f_X(z)
\end{aligned}$$

**Ex 3.**  $X$  has pdf  $f(x) = \begin{cases} 2x, & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$

Find the distribution of  $Y = 8X^3$ .

*Answer:* First, note that  $0 < y < 8$ .

$$\begin{aligned}
F_Y(y) &= P(Y \leq y) = P(8X^3 \leq y) = P\left\{X \leq \left(\frac{y}{8}\right)^{1/3}\right\} = \int_0^{\left(\frac{y}{8}\right)^{1/3}} 2x \, dx = \frac{1}{4}y^{2/3}, \quad 0 < y < 8 \\
f_Y(y) &= \frac{\partial}{\partial y} F_Y(y) = \begin{cases} \frac{1}{6\sqrt[3]{y}}, & 0 < y < 8 \\ 0, & \text{otherwise} \end{cases}
\end{aligned}$$

## PDF Method

Basically same as the cdf method, except that the pdf is used instead of the cdf. Suppose  $Y = T(X)$  with a transformation,  $T$ : 1-1 transformation from  $\mathcal{A}$  onto  $\mathcal{B}$ . Then  $Y$  will have a pdf:

$$f_Y(y) = \begin{cases} f_X\{T^{-1}(y)\} |J_{T^{-1}(y)}|, & y \in \mathcal{B} \\ 0, & \text{otherwise} \end{cases}$$

**Ex 4.**  $X$  has pdf  $f(x) = \begin{cases} 2x, & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$

Find the distribution of  $Y = 8X^3$ .

*Answer:* First, note that  $0 < y < 8$ .

$$\begin{aligned}
T : y &= 8x^3 \\
T^{-1} : x &= \left(\frac{y}{8}\right)^{1/3} = \frac{1}{2}y^{1/3} \\
f_X\{T^{-1}(y)\} &= 2 \cdot \frac{1}{2}y^{1/3} = y^{1/3} \\
|J_{T^{-1}(y)}| &= \left|\frac{\partial}{\partial y} \left(\frac{1}{2}y^{1/3}\right)\right| = \frac{1}{6}y^{-2/3}
\end{aligned}$$

$$\therefore f_Y(y) = f_X \{T^{-1}(y)\} |J_{T^{-1}(y)}| = \frac{1}{6}y^{-1/3}, 0 < y < 8$$

**Ex 5.**  $X \sim \text{uniform}(0, 1)$ . Find the distribution of  $Y = -2 \ln X$ .

*Answer:* First, note that  $y > 0$ .

$$T : y = -2 \ln x \quad : \text{1-1 from } \mathcal{A} = (0, 1) \text{ onto } \{y : y > 0\}$$

$$T^{-1} : x = e^{-\frac{1}{2}y}$$

$$f_X \{T^{-1}(y)\} = 1, \quad y > 0$$

$$|J_{T^{-1}(y)}| = \left| \frac{\partial}{\partial y} \left( e^{-\frac{1}{2}y} \right) \right| = \frac{1}{2}e^{-\frac{1}{2}y}$$

$$\therefore f_Y(y) = f_X \{T^{-1}(y)\} |J_{T^{-1}(y)}| = \frac{1}{2}e^{-\frac{1}{2}y}, y > 0$$

$$\text{i.e., } Y \sim \chi^2_{(df=2)}.$$

**Ex 6.** Let  $X_1 \sim \Gamma(\alpha, 1)$ ,  $X_2 \sim \Gamma(\beta, 1)$  and  $X_1, X_2$ : independent. Show

1. distribution of  $\frac{X_1}{X_1 + X_2}$  is  $\beta(\alpha, \beta)$
2.  $X_1 + X_2 \sim \Gamma(\alpha + \beta, 1)$
3.  $X_1 + X_2$  and  $\frac{X_1}{X_1 + X_2}$  are independent.

*Proof:*

$$T : \begin{cases} y_1 = x_1 + x_2 \\ y_2 = \frac{x_1}{x_1 + x_2} \end{cases} \quad : \text{1-1 from } \mathcal{A} = (0, \infty) \times (0, \infty) \text{ onto } \{y_1 y_2 > 0, y_1 - y_1 y_2 > 0\}$$

$$T^{-1} : \begin{cases} x_1 = y_1 y_2 \\ x_2 = y_1 - y_1 y_2 \end{cases} \quad y_1 > 0, 0 < y_2 < 1$$

$$|J_{T^{-1}}| = \begin{vmatrix} y_2 & y_1 \\ 1 - y_2 & -y_1 \end{vmatrix} = |y_1| = y_1$$

$$f(x_1, x_2) = \frac{1}{\Gamma(\alpha)\Gamma(\beta)} x_1^{\alpha-1} x_2^{\beta-1} e^{-x_1} e^{-x_2}$$

$$\therefore f(y_1, y_2) = \frac{1}{\Gamma(\alpha)\Gamma(\beta)} (y_1 y_2)^{\alpha-1} \{y_1(1 - y_2)\}^{\beta-1} e^{-y_1} \cdot y_1, \quad y_1 > 0, 0 < y_2 < 1$$

$$= \frac{1}{\Gamma(\alpha)\Gamma(\beta)} y_1^{\alpha+\beta-1} e^{-y_1} y_2^{\alpha-1} (1 - y_2)^{\beta-1}$$

$$= \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} y_2^{\alpha-1} (1 - y_2)^{\beta-1} I_{(0,1)}(y_2) \cdot \frac{1}{\Gamma(\alpha + \beta)} y_1^{\alpha+\beta-1} e^{-y_1} I_{(0,\infty)}(y_1)$$

$$= f(y_2) \cdot f(y_1), \quad Y_2 \sim \beta(\alpha, \beta) \text{ and } Y_1 \sim \Gamma(\alpha + \beta, 1)$$

**Ex 7.** Let  $X_1 \sim \Gamma(\alpha, \theta)$ ,  $X_2 \sim \Gamma(\beta, \theta)$  and  $X_1, X_2$ : independent. Show

1. distribution of  $\frac{X_1}{X_1 + X_2}$  is  $\beta(\alpha, \beta)$
2.  $X_1 + X_2 \sim \Gamma(\alpha + \beta, \theta)$
3.  $X_1 + X_2$  and  $\frac{X_1}{X_1 + X_2}$  are independent.

*Proof:*

$$T : \begin{cases} y_1 = x_1 + x_2 \\ y_2 = \frac{x_1}{x_1 + x_2} \end{cases} : 1-1 \text{ from } \mathcal{A} = (0, \infty) \times (0, \infty) \text{ onto } \{y_1 y_2 > 0, y_1 - y_1 y_2 > 0\}$$

$$T^{-1} : \begin{cases} x_1 = y_1 y_2 \\ x_2 = y_1 - y_1 y_2 \end{cases} \quad y_1 > 0, 0 < y_2 < 1$$

$$|J_{T^{-1}}| = \begin{vmatrix} y_2 & y_1 \\ 1 - y_2 & -y_1 \end{vmatrix} = |y_1| = y_1$$

$$f(x_1, x_2) = \frac{1}{\Gamma(\alpha)\Gamma(\beta)\theta^\alpha\theta^\beta} x_1^{\alpha-1} x_2^{\beta-1} e^{-x_1/\theta} e^{-x_2/\theta}$$

$$\begin{aligned} \therefore f(y_1, y_2) &= \frac{1}{\Gamma(\alpha)\Gamma(\beta)\theta^\alpha\theta^\beta} (y_1 y_2)^{\alpha-1} \{y_1(1 - y_2)\}^{\beta-1} e^{-y_1/\theta} \cdot y_1, \quad y_1 > 0, 0 < y_2 < 1 \\ &= \frac{1}{\Gamma(\alpha)\Gamma(\beta)\theta^\alpha\theta^\beta} y_1^{\alpha+\beta-1} e^{-y_1/\theta} y_2^{\alpha-1} (1 - y_2)^{\beta-1} \\ &= \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} y_2^{\alpha-1} (1 - y_2)^{\beta-1} I_{(0,1)}(y_2) \cdot \frac{1}{\Gamma(\alpha + \beta)\theta^{\alpha+\beta}} y_1^{\alpha+\beta-1} e^{-y_1/\theta} I_{(0,\infty)}(y_1) \\ &= f(y_2) \cdot f(y_1), \quad Y_2 \sim \beta(\alpha, \beta) \text{ and } Y_1 \sim \Gamma(\alpha + \beta, \theta) \end{aligned}$$

**Corollary**  $X_1 \sim \chi^2(\alpha)$ ,  $X_2 \sim \chi^2(\beta)$  and  $X_1, X_2$ : independent. Let  $Y_1 = \frac{X_1}{X_1 + X_2}$  and  $Y_2 = \frac{X_2}{X_1 + X_2}$ . Then,  $Y_1 \sim \beta\left(\frac{\alpha}{2}, \frac{\beta}{2}\right)$ ,  $Y_2 \sim \Gamma\left(\frac{\alpha+\beta}{2}, 2\right) \sim \chi^2(\alpha + \beta)$  and  $Y_1, Y_2$  are independent.

**Ex 8.** Let  $X_1, X_2$ : iid exponential with  $\lambda$ , i.e.,  $\Gamma\left(1, \frac{1}{\lambda}\right)$ . Find the distribution of  $Y_1 = X_1 - X_2$ .

Answer:

$$\begin{aligned}
T &: \begin{cases} y_1 = x_1 - x_2 \\ y_2 = x_2 \end{cases} : \text{1-1 from } \mathcal{A} = (0, \infty) \times (0, \infty) \text{ onto } \{y_1 + y_2 > 0, y_2 > 0\} \\
T^{-1} &: \begin{cases} x_1 = y_1 + y_2 \\ x_2 = y_2 \end{cases} \quad y_1 + y_2 > 0, y_2 > 0 \\
|J_{T^{-1}}| &= \begin{vmatrix} 1 & 1 \\ 0 & 1 \end{vmatrix} = 1 \\
f(x_1, x_2) &= \lambda^2 e^{-\lambda(x_1+x_2)}, \quad x_1 > 0, x_2 > 0 \\
\therefore f(y_1, y_2) &= \lambda^2 e^{-\lambda(y_1+2y_2)}, \quad y_1 + y_2 > 0, y_2 > 0 \\
\therefore f_1(y_1) &= \int_{-\infty}^{\infty} f(y_1, y_2) dy_2 \\
&= \begin{cases} \int_0^{\infty} \lambda^2 e^{-\lambda(y_1+2y_2)} dy_2, & y_1 > 0 \\ \int_{-y_1}^0 \lambda^2 e^{-\lambda(y_1+2y_2)} dy_2, & y_1 \leq 0 \end{cases} \\
&= \begin{cases} \frac{\lambda}{2} e^{-\lambda y_1}, & y_1 > 0 \\ \frac{\lambda}{2} e^{\lambda y_1}, & y_1 \leq 0 \end{cases} \\
&= \frac{\lambda}{2} e^{-\lambda|y_1|}, \quad -\infty < y_1 < \infty
\end{aligned}$$

$Y_1$  is called the double-exponential distribution or Lapace distribution.

**Ex 9. Box-Muller transformation:** Let  $X_1, X_2$ : iid uniform (0,1). Then

$$\begin{aligned}
Y_1 &= \sqrt{-2 \ln X_1} \cos(2\pi X_2) \\
Y_2 &= \sqrt{-2 \ln X_1} \sin(2\pi X_2) \\
\Rightarrow Y_1, Y_2 &\sim iid N(0, 1)
\end{aligned}$$

*Proof:*

$$\begin{aligned}
T &: \begin{cases} y_1 = \sqrt{-2 \ln x_1} \cos(2\pi x_2) \\ y_2 = \sqrt{-2 \ln x_1} \sin(2\pi x_2) \end{cases} : \text{1-1 from } \mathcal{A} = (0, 1) \times (0, 1) \text{ onto } \mathbb{R}^2 \\
T^{-1} &: \begin{cases} x_1 = \exp\left(-\frac{y_1^2 + y_2^2}{2}\right) \\ x_2 = \frac{1}{2\pi} \tan^{-1}\left(\frac{y_2}{y_1}\right) \end{cases} \\
|J_{T^{-1}}| &= \begin{vmatrix} -x_1 e^{-\frac{y_1^2 + y_2^2}{2}} & -x_2 e^{-\frac{y_1^2 + y_2^2}{2}} \\ \frac{1}{2\pi} \frac{-y_2/y_1^2}{1+(y_2/y_1)^2} & \frac{1}{2\pi} \frac{1/y_1}{1+(y_2/y_1)^2} \end{vmatrix} = \frac{1}{2\pi} e^{-\frac{y_1^2 + y_2^2}{2}} \\
f(x_1, x_2) &= I_{(0,1)}(x_1) \cdot I_{(0,1)}(x_2) \\
\therefore f(y_1, y_2) &= \frac{1}{2\pi} e^{-\frac{y_1^2 + y_2^2}{2}}, \quad -\infty < y_1, y_2 < \infty \\
\therefore Y_1, Y_2 &\sim \text{iid } N(0, 1)
\end{aligned}$$

**Ex 10.**  $X_1, X_2$ : iid  $N(0, 1)$ . Then  $Y = \frac{X_1}{X_2} \sim \mathcal{C}(0, 1)$ , i.e., Cauchy with pdf  $\frac{1}{\pi} \frac{1}{1+y^2}$ ,  $-\infty < y < \infty$ .

*Proof:*

$$\begin{aligned}
T &: \begin{cases} y_1 = \frac{x_1}{x_2} \\ y_2 = x_2 \end{cases} : \text{1-1 from } \mathbb{R}^2 \setminus \{x_2 = 0\} \text{ onto } \mathbb{R}^2 \\
T^{-1} &: \begin{cases} x_1 = y_1 y_2 \\ x_2 = y_2 \end{cases} \quad y_1 + y_2 > 0, y_2 > 0 \\
|J_{T^{-1}}| &= \begin{vmatrix} y_2 & y_1 \\ 0 & 1 \end{vmatrix} = |y_2| \\
f(x_1, x_2) &= \frac{1}{2\pi} e^{-\frac{x_1^2 + x_2^2}{2}}, \quad x_1, x_2 \in \mathbb{R} \\
\therefore f(y_1, y_2) &= \frac{1}{2\pi} |y_2| e^{-\frac{y_1^2 y_2^2 + y_2^2}{2}}, \quad y_1, y_2 \in \mathbb{R} \\
\therefore f_1(y_1) &= \int_{-\infty}^{\infty} f(y_1, y_2) dy_2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} |y_2| e^{-\frac{1+y_1^2}{2} y_2^2} dy_2 \\
&= \frac{1}{\pi} \int_0^{\infty} y_2 e^{-(1+y_1^2)t} dt = \frac{1}{\pi} \frac{1}{1+y_1^2}, \quad y_1 \in \mathbb{R}
\end{aligned}$$

**One note:** General form of the Cauchy pdf:

$$X \sim \mathcal{C}(\mu, \sigma) \quad \rightarrow \quad f(x) = \frac{1}{\sigma} \frac{1}{\pi} \frac{1}{1 + \left(\frac{x-\mu}{\sigma}\right)^2}, \quad x \in \mathbb{R}$$

## Selected Problems

1. Let  $X$  have the pdf  $f(x) = 4x^3$ ,  $0 < x < 1$ . Find the pdf of  $Y = X^2$ .
2. Let  $X$  have the pdf  $f(x) = xe^{-x^2/2}$ ,  $0 < x < \infty$ . Find the pdf of  $Y = X^2$ .
3. Let  $X$  have a  $\Gamma(\alpha = 3, \theta = 2)$ . Determine the pdf of  $Y = \sqrt{X}$ .
4. Let  $X$  have the pdf  $f(x) = 2x$ ,  $0 < x < 1$ .
  - (a) Find the cdf of  $X$ .
  - (b) Describe how an observation of  $X$  can be simulated.
  - (c) Simulate 10 observations of  $X$ .
5. Let  $X$  have the pdf  $f(x) = \theta x^{\theta-1}$ ,  $0 < x < 1$ ,  $0 < \theta < \infty$ . Find the distribution of  $Y = -2\theta \ln X$ .
6. Let  $X$  have a *logistic distribution* with the pdf  $f(x) = \frac{e^{-x}}{(1 + e^{-x})^2}$ ,  $-\infty < x < \infty$ . Show that  $Y = \frac{1}{1 + e^{-X}}$  has a uniform  $(0, 1)$  distribution.
7. Suppose the lifetime (in years) of a product is  $Y = 5X^{0.7}$ , where  $X$  has an exponential distribution with mean 1. Find the cdf and pdf of  $Y$ .