Lesson 1.1

The Cumulative Distribution Function Technique

Introduction

The objective of statistics is to make inferences about a population based on information contained in a sample taken from that population. The quantities used to estimate population parameters or to make decisions about a population are functions of the n random observations that appear in a sample.

To illustrate, consider the problem of estimating a population mean, μ . Intuitively, we draw a random sample of n observations, y_1, y_2, \dots, y_n , from the population and use the sample mean

$$\bar{Y} = \frac{\sum_{i=1}^{n} Y_i}{n}$$

to estimate μ .

If the probability distributuion of the $Y_i's$ is known what would be the distribution of the sample mean \bar{Y} . If $Y_i's$ are random variables then \bar{Y} is likewise a random variable.

In this module, we shall study methods in finding the distribution of functions of random variables. These methods include:

- 1. Cumulative Distribution Function (CDF) Technique
- 2. Transformation Technique
- 3. Moment Generation Function (MGF) Technique

The Cumulative Distribution Function (CDF) Technique

Suppose Y is a continuous random variable with cumulative distribution function $F_Y(y) = P(Y \leq y)$. The CDF technique is useful if $F_Y(y)$ has a closed form. The steps in the CDF method are:

- 1. Find a closed form of the CDF of Y.
- 2. Find the support of U = g(Y).
- 3. Write $F_U(u) = P(U \le U)$ in terms of F(Y).
- 4. Differentiate $F_U(u)$ to obtain the PDF of U.

Example 1.1.1

Suppose $Y \sim U(0,1)$. Find the distribution of U = g(Y) = -lnY.

SOLUTION

Verify that the CDF of Y is given by

$$F_Y(y) = \begin{cases} 0, \ y \le 0 \\ y, \ 0 < y < 1 \\ 1, \ y \ge 1 \end{cases}$$

Since the support for $Y \sim U(0,1)$ is $R_Y = \{y: 0 < y < 1\}$, thus, it follows that the support for U = -lnY is $R_U = \{u: u > 0\}$.

Now, we are going to find the CDF of U.

$$\begin{split} F_U(u) &= P(U \le u) \\ &= P(-lnY \le u) \\ &= P(lnY > -u) \\ &= P(Y > e^{-u}) \\ &= 1 - P(Y \le e^{-u}) \\ &= 1 - F_Y(e^{-u}) \end{split}$$

Since $F_Y(y) = y$ for 0 < y < 1, hence, for u > 0 we have

$$\begin{split} F_U(u) &= 1 - F_Y(e^{-u}) \\ &= 1 - e^{-u} \end{split}$$

To get the PDF of U, we have

$$f_U(u) = \frac{d}{du} F_U(u)$$
$$= \frac{d}{du} [1 - e^{-u}]$$
$$= e^{-u}$$

Therefore,

$$f_U(u) = \begin{cases} e^{-u}, \ u > 0 \\ 0, \text{ elsewhere} \end{cases}$$

which shows that $U \sim Exp(1)$ which is an exponential distribution with mean $\beta = 1$.

Example 1.1.2

A process for refining sugar yields up to 1 ton of pure sugar per day, but the actual amount produced, Y, is a random variable because of machine breakdowns and other slowdowns. Suppose that Y has density function given by

$$f_Y(y) = \begin{cases} 2y, \ 0 \le y \le 1 \\ 0, \ \text{elsewhere} \end{cases}$$

The company is paid at the rate of \$300 per ton for the refined sugar, but it also has a fixed overhead cost of \$100 per day. Thus the daily profit, in hundreds of dollars, is U = 3Y - 1. Find the probability density function for U.

SOLUTION

Verify that the CDF of Y is given by

$$F_Y(y) = \begin{cases} 0, \ y < 0 \\ y^2, \ 0 \le y < 1 \\ 1, \ y > 1 \end{cases}$$

Note that the support of U=3Y-1 is $R_U=\{u:-1\leq u\leq 2\}.$

The CDF of U is obtained as follows:

$$\begin{split} F_U(u) &= P(U \leq u) \\ &= P(3Y-1 \leq u) \\ &= P(Y \leq \frac{u+1}{3}) \\ &= F_Y\left(\frac{u+1}{3}\right) \end{split}$$

That is, for $-1 \le u \le 2$,

$$F_U(u) = F_Y\left(\frac{u+1}{3}\right)$$
$$= \left(\frac{u+1}{3}\right)^2$$

And the density function for U is

$$\begin{split} f_U(u) &= \frac{d}{du} F_U(u) \\ &= \frac{d}{du} \left[\left(\frac{u+1}{3} \right)^2 \right] \\ &= \frac{2}{9} (u+1) \end{split}$$

Therefore,

$$f_U(u) = \begin{cases} \frac{2}{9}(u+1), \ -1 \leq u \leq 2 \\ 0, \text{ elsewhere} \end{cases}$$

Example 1.1.3

Let Y be a random variable with PDF

$$f_Y(y) = \begin{cases} \frac{3}{2}y^2, \, -1 \leq y \leq 1 \\ 0, \, \text{elsewhere} \end{cases}$$

SOLUTION [Left as a classroom exercise!]