

## MODULE 7

### LIMITING DISTRIBUTIONS

#### LECTURES 37-42

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## MODULE 7

### LIMITING DISTRIBUTIONS

#### LECTURE 37

#### Topics

##### 7.1 CONVERGENCE IN DISTRIBUTION AND PROBABILITY

Let  $\underline{T} = (T_1, \dots, T_n)$  be a random vector having a probability density function/probability mass function (p.d.f./p.m.f.)  $f_{\underline{T}}(\cdot)$  and let  $h: \mathbb{R}^n \rightarrow \mathbb{R}$  be a Borel function. Suppose that the distribution of random variable  $X_n = h(\underline{T})$  is desired. Very often it is not possible to derive the expression for distribution (i.e., p.d.f. or p.m.f.) of  $X_n = h(\underline{T})$ . To make this point clear let  $T_1, \dots, T_n$  be a random sample from  $\text{Be}(a, b)$  distribution, where  $a$  and  $b$  are

positive real constants, and suppose that the distribution (i.e., the distribution function or a p.d.f.) of the sample mean  $\bar{T}_n = \frac{1}{n} \sum_{i=1}^n T_i$  is desired. The form of the p.d.f. (or distribution function) of  $\bar{T}_n$  is so complicated (it involves multiple integrals which cannot be expressed in a closed form) that hardly anybody would be interested in using it. Therefore, it will be helpful if we can approximate the distribution of  $\bar{T}_n$  by a distribution which is mathematically tractable. In this module we will develop a theory which will help us in approximating distributions of a sequence  $\{X_n\}_{n \geq 1}$  of random variables for large values of  $n$  (say, as  $n \rightarrow \infty$ ). Such approximations are quite useful in statistical inference problems.

## 7.1 CONVERGENCE IN DISTRIBUTION AND PROBABILITY

Let  $\{X_n\}_{n \geq 1}$  be a sequence of random variables with corresponding sequence of distribution functions (d.f.s) as  $\{F_n\}_{n \geq 1}$ . Suppose that an approximation to the distribution of  $X_n$  (i.e., of  $F_n$ ) is desired, for large values of  $n$  (say, as  $n \rightarrow \infty$ ). It may be tempting to approximate  $F_n(\cdot)$  by  $F(x) = \lim_{n \rightarrow \infty} F_n(x), x \in \mathbb{R}$ . However, as the following examples illustrate,  $F(x) = \lim_{n \rightarrow \infty} F_n(x), x \in \mathbb{R}$ , may not be a d.f..

### Example 7.1

- (i) Let  $\{X_n\}_{n \geq 1}$  be sequence of random variables with  $P(\{X_n = n\}) = 1, n = 1, 2, \dots$ . Then the d.f. of  $X_n$  is given by

$$F_n(x) = \begin{cases} 0, & \text{if } x < n \\ 1, & \text{if } x \geq n \end{cases}, \quad n = 1, 2, \dots$$

We have  $F(x) \stackrel{\text{def}}{=} \lim_{n \rightarrow \infty} F_n(x) = 0, \forall x \in \mathbb{R}$ . Clearly  $F$  is not a d.f..

- (ii) Let  $X_n \sim U(-n, n), n = 1, 2, \dots$ . Then the d.f. of  $X_n$  is

$$F_n(x) = \begin{cases} 0, & \text{if } x < -n \\ \frac{x+n}{2n}, & \text{if } -n \leq x < n, \\ 1, & \text{if } x \geq n \end{cases}, \quad n = 1, 2, \dots$$

Clearly  $F(x) \stackrel{\text{def}}{=} \lim_{n \rightarrow \infty} F_n(x) = \frac{1}{2}, \forall x \in \mathbb{R}$  and  $F(\cdot)$  is not a d.f. ■

The above examples illustrate that a sequence  $\{F_n\}_{n \geq 1}$  of d.f.s on  $\mathbb{R}$  may converge, at all points, but the limiting function  $F(x) = \lim_{n \rightarrow \infty} F_n(x), x \in \mathbb{R}$ , may not be a d.f..

The following example illustrates that if a sequence  $\{F_n\}_{n \geq 1}$  of d.f.s converges at every point then it may be too restrictive to require that  $\{F_n\}_{n \geq 1}$  converges to a d.f.  $F$  at all points (i.e., to require that  $\lim_{n \rightarrow \infty} F_n(x) = F(x), \forall x \in \mathbb{R}$ , for some d.f.  $F$ ).

**Example 7.2**

Let  $\{X_n\}_{n \geq 1}$  be a sequence of random variables with  $P\left(\left\{X_n = \frac{1}{n}\right\}\right) = 1, n = 1, 2, \dots$ . Then the d.f. of  $X_n$  is

$$F_n(x) = \begin{cases} 0, & \text{if } x < \frac{1}{n} \\ 1, & \text{if } x \geq \frac{1}{n} \end{cases}, \quad n = 1, 2, \dots$$

Clearly,

$$F(x) \stackrel{\text{def}}{=} \lim_{n \rightarrow \infty} F_n(x) = \begin{cases} 0, & \text{if } x \leq 0 \\ 1, & \text{if } x > 0 \end{cases}$$

is not a d.f. (it is not right continuous at  $x = 0$ ). However,  $F$  can be converted into a distribution function

$$F^*(x) = \begin{cases} 0, & \text{if } x < 0 \\ 1, & \text{if } x \geq 0 \end{cases}$$

by changing its value at the point 0 (the point of discontinuity of  $F$ ). Since  $P\left(\left\{X_n = \frac{1}{n}\right\}\right) = 1, n = 1, 2, \dots$ , and  $\lim_{n \rightarrow \infty} \frac{1}{n} = 0$ , a natural approximation of  $F_n$  seems to be the distribution function of a random variable  $X$  that is degenerate at 0 (i.e.,  $P(\{X = 0\}) = 1$ ). Note that  $F^*$  is the d.f. of random variables  $X$  that is degenerate at 0. The above discussion suggests that it is too restrictive to require

$$\lim_{n \rightarrow \infty} F_n(x) = F^*(x), \forall x \in \mathbb{R},$$

and that exceptions may be permitted at the points of discontinuities of  $F^*$ . ■

**Definition 1.1**

Let  $\{X_n\}_{n \geq 1}$  be a sequence of random variables and let  $F_n$  be the d.f. of  $X_n, n = 1, 2, \dots$

- (i) Let  $X$  be a random variable with d.f.  $F$ . The sequence  $\{X_n\}_{n \geq 1}$  is said to *converge in distribution* to  $X$ , as  $n \rightarrow \infty$  (written as  $X_n \xrightarrow{d} X$ , as  $n \rightarrow \infty$ ) if  $\lim_{n \rightarrow \infty} F_n(x) = F(x), \forall x \in C_F$ , where  $C_F$  is the set of continuity points of  $F$ . The d.f.  $F$  (or the corresponding p.d.f/p.m.f.) is called the *limiting distribution* of  $X_n$ , as  $n \rightarrow \infty$ .
- (ii) Let  $c \in \mathbb{R}$ . The sequence  $\{X_n\}_{n \geq 1}$  is said to converge in probability to  $c$ , as  $n \rightarrow \infty$  (written as  $X_n \xrightarrow{p} c$ , as  $n \rightarrow \infty$ ) if  $X_n \xrightarrow{d} X$ , as  $n \rightarrow \infty$ , where  $X$  is a random variable that is degenerate at  $c$ . ■

**Remark 1.1**

- (i) Suppose that  $X_n \xrightarrow{d} X$ , as  $n \rightarrow \infty$ . Since the set  $D_F = C_F^c = \mathbb{R} - C_F$  of discontinuity points of limiting d.f.  $F$  is at most countable we have  $\lim_{n \rightarrow \infty} F_n(x) = F(x)$  everywhere except, possibly, at a countable number of points.
- (ii) Note that the distribution function of a random variable degenerate at point  $c \in \mathbb{R}$  is given by

$$F(x) = \begin{cases} 0, & \text{if } x < c \\ 1, & \text{if } x \geq c \end{cases}.$$

Thus we have

$$X_n \xrightarrow{p} c, \text{ as } n \rightarrow \infty \Leftrightarrow \lim_{n \rightarrow \infty} F_n(x) = \begin{cases} 0, & \text{if } x < c \\ 1, & \text{if } x \geq c \end{cases}.$$

- (iii) Suppose that  $X_n \xrightarrow{d} X$ , as  $n \rightarrow \infty$ . If the random variable  $X$  is of continuous type (i.e.,  $C_F = \mathbb{R}$ ) then  $\lim_{n \rightarrow \infty} F_n(x) = F(x), \forall x \in \mathbb{R}$ .
- (iv) Note that, for a real constant  $c$ ,  $X_n \xrightarrow{p} c$  if, and only if,  $X_n - c \xrightarrow{p} 0$ , as  $n \rightarrow \infty$ . ■

**Example 1.3**

Let  $\{X_n\}_{n \geq 1}$  be a sequence of random variables such that  $P(\{X_n = 0\}) = \frac{1}{n} = 1 - P\left(\left\{X_n = \frac{1}{n}\right\}\right)$ ,  $n = 1, 2, \dots$ . Show that  $X_n \xrightarrow{p} 0$ , as  $n \rightarrow \infty$ .

**Solution.** Let  $F$  be the d.f. of a random variable degenerate at 0, i.e.,

$$F(x) = \begin{cases} 0, & \text{if } x < 0 \\ 1, & \text{if } x \geq 0 \end{cases}.$$

Since  $F$  is continuous everywhere except at point 0 (i.e.,  $C_F = \mathbb{R} - \{0\}$ ), we need to show that  $\lim_{n \rightarrow \infty} F_n(x) = F(x), \forall x \in \mathbb{R} - \{0\}$ , where  $F_n(\cdot)$  is the d.f. of  $X_n, n = 1, 2, \dots$

We have

$$F_n(x) = \begin{cases} 0, & \text{if } x < 0 \\ \frac{1}{n}, & \text{if } 0 \leq x < \frac{1}{n} \\ 1, & \text{if } x \geq \frac{1}{n} \end{cases}, \quad n = 1, 2, \dots$$

$$\xrightarrow{n \rightarrow \infty} \begin{cases} 0, & \text{if } x \leq 0 \\ 1, & \text{if } x > 0 \end{cases}.$$

Clearly  $\lim_{n \rightarrow \infty} F_n(x) = F(x), \forall x \in \mathbb{R} - \{0\}$ . ■

### Example 1.4

Let  $X_1, X_2, \dots$  be a sequence of independent and identically distributed (i.i.d.)  $U(0, \theta)$  random variables, where  $\theta > 0$ . Let  $X_{n:n} = \max\{X_1, \dots, X_n\}$  and let  $Y_n = n(\theta - X_{n:n})$ ,  $n = 1, 2, \dots$

- (i) Show that  $X_{n:n} \xrightarrow{p} \theta$ , as  $n \rightarrow \infty$ ;
- (ii) Find the limiting distribution of  $\{Y_n\}_{n \geq 1}$ .

**Solution.**

- (i) Let  $H_n$  be the d.f. of  $X_{n:n}$ ,  $n = 1, 2, \dots$ , and let

$$H(x) = \begin{cases} 0, & \text{if } x < \theta \\ 1, & \text{if } x \geq \theta \end{cases}$$

be the d.f. of random variable degenerate at  $\theta$ . We need to show that  $\lim_{n \rightarrow \infty} H_n(x) = H(x), \forall x \in \mathbb{R} - \{\theta\}$ .

We have, for  $x \in \mathbb{R}$ ,

$$\begin{aligned} H_n(x) &= P(\{X_{n:n} \leq x\}) \\ &= P(\{\max\{X_1, \dots, X_n\} \leq x\}) \\ &= P(\{X_i \leq x, i = 1, \dots, n\}) \\ &= \prod_{i=1}^n P(\{X_i \leq x\}) \quad (\text{since } X_i\text{s are independent}) \\ &= [F(x)]^n, n = 1, 2, \dots, \quad (\text{since } X_i\text{s are identically distributed}), \end{aligned}$$

where

$$F(x) = \begin{cases} 0, & \text{if } x < 0 \\ \frac{x}{\theta}, & \text{if } 0 \leq x < \theta \\ 1, & \text{if } x \geq \theta \end{cases}$$

is the common distribution function of  $X_1, X_2, \dots$

Thus

$$\begin{aligned}
H_n(x) &= \begin{cases} 0, & \text{if } x < 0 \\ \left(\frac{x}{\theta}\right)^n, & \text{if } 0 \leq x < \theta \\ 1, & \text{if } x \geq \theta \end{cases} \\
&\xrightarrow{n \rightarrow \infty} \begin{cases} 0, & \text{if } x < \theta \\ 1, & \text{if } x \geq \theta \end{cases} \\
&= H(x), \forall x \in \mathbb{R}.
\end{aligned}$$

(ii) For  $y \in \mathbb{R}$ , we have

$$\begin{aligned}
F_{Y_n}(y) &= P(\{Y_n \leq y\}) \\
&= P\left(\left\{X_{n:n} \geq \theta - \frac{y}{n}\right\}\right) \\
&= 1 - H_n\left(\left(\theta - \frac{y}{n}\right) -\right) \\
&= 1 - H_n\left(\theta - \frac{y}{n}\right) \quad (\text{since } H_n \text{ is continuous}) \\
&= \begin{cases} 0, & \text{if } y \leq 0 \\ 1 - \left(1 - \frac{y}{n\theta}\right)^n, & \text{if } 0 < y \leq n\theta, \quad n = 1, 2, \dots \\ 1, & \text{if } y > n\theta \end{cases} \\
&\xrightarrow{n \rightarrow \infty} \begin{cases} 0, & \text{if } y \leq 0 \\ 1 - e^{-\frac{y}{\theta}}, & \text{if } y > 0 \end{cases} \\
&= G(y), \text{ say.}
\end{aligned}$$

Note that  $G(\cdot)$  is the d.f. of  $\text{Exp}(\theta)$  random variable. Thus  $Y_n \xrightarrow{d} Y \sim \text{Exp}(\theta)$ , as  $n \rightarrow \infty$ . ■

In the above example we saw that  $X_{n:n} \xrightarrow{p} \theta$ , as  $n \rightarrow \infty$ , and  $n(\theta - X_{n:n}) \xrightarrow{d} Y \sim \text{Exp}(\theta)$ , as  $n \rightarrow \infty$ , i.e., the limiting distribution of  $X_{n:n}$  is degenerate (at  $\theta$ ) and, to get a non-degenerate limiting distribution, we needed normalized version  $Y_n = n(\theta - X_{n:n})$  of  $X_{n:n}$ ,  $n = 1, 2, \dots$ . This phenomenon is observed quite commonly. Generally, we will have a sequence  $\{X_n\}_{n \geq 1}$  of random variables, such that  $X_n \xrightarrow{p} c$ , as  $n \rightarrow \infty$  for some real constant  $c$  (i.e., the limiting distribution of  $X_n$  is degenerate at  $c$ ). In order to get a non-degenerate limiting distribution a normalized version  $Z_n = n^r(X_n - c)$  (or  $Z_n = n^r(c - X_n)$ ),  $r > 0$ , of  $X_n$ ,  $n = 1, 2, \dots$  is considered. Typically there is a choice of  $r > 0$  such that the limiting distribution of  $Z_n$  is non-degenerate.