# CSIR-UGC NET-June 2013-Problem(72)

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### Convergence in Probability

A sequence of random variables  $X_1, X_2, X_3, \ldots$  converges in probability to a random variable X, if

$$\lim_{n \to \infty} \Pr(|X_n - X| \ge \epsilon) = 0, \forall \epsilon > 0$$
 (1)

Notation :  $X_n \xrightarrow{p} X$ 

### Convergence in Distribution

A sequence of random variables  $X_1, X_2, X_3, \ldots$  converges in distribution to a random variable X, if

$$\lim_{n \to \infty} F_{X_n}(x) = F_X(x) \tag{2}$$

for all x at which  $F_X(x)$  is continuous.

Notation :  $X_n \xrightarrow{d} X$ 

### Random Sampling

A collection of random variables  $X_1, X_2, \ldots, X_n$  is said to be a random sample of size n if they are independent and identically distributed, i.e,

- $\bigcirc$   $X_1, X_2, \dots, X_n$  are independent random variables
- ② They have the same distribution (Let us denote it by  $F_X(x)$ ), i.e,

$$F_X(x) = F_{X_1}(x) = F_{X_2}(x) = \dots = F_{X_n}(x), \forall x \in \mathbb{R}$$
 (3)

#### **Order Statistics**

Given a random sample  $X_1, X_2, \ldots, X_n$ , the sequence  $X_{(1)}, X_{(2)}, \ldots, X_{(n)}$  is called the order statistics of it. Here,

$$X_{(1)} = \min(X_1, X_2, \dots, X_n)$$
 (4)

$$X_{(2)} = \text{the } 2^{nd} \text{ smallest of } X_1, X_2, \dots, X_n$$
 (5)

$$X_{(n)} = \max(X_1, X_2, \dots, X_n)$$
 (7)

#### Distribution of the maximum

Let's calculate the CDF, PDF of  $X_{(n)}$ 

$$F_{X_{(n)}}(x) = \Pr(X_{(n)} \le x) = \Pr(X_1 \le x, X_2 \le x, \dots, X_n \le x)$$
(8)  
= 
$$\Pr(X_1 \le x) \Pr(X_2 \le x) \dots \Pr(X_n \le x) (\because \text{ independence})$$
(9)  
= 
$$[\Pr(X_1 \le x)]^n (\because \text{ identical distribution}) = [F_X(x)]^n$$
(10)

$$f_{X_{(n)}}(x) = \frac{d}{dx} \left( F_{X_{(n)}}(x) \right) = \frac{d}{dx} \left( \left[ F_X(x) \right]^n \right) \tag{11}$$

$$= n \left( \left[ F_X(x) \right]^{n-1} \right) \frac{d}{dx} \left( F_X(x) \right) \tag{12}$$

$$= n \left[ F_X(x) \right]^{n-1} f_X(x) \left( \because \frac{d}{dx} \left( F_X(x) \right) = f_X(x) \right) \tag{13}$$

#### Uniform Distribution

A continuous random variable X is said to have a Uniform Distribution over the interval (a,b), shown as  $X \sim \textit{Uniform}(a,b)$ , if,

$$f_X(x) = \begin{cases} \frac{1}{b-a}, & a < x < b \\ 0, & otherwise \end{cases}, F_X(x) = \begin{cases} x, & a < x < b \\ 1, & x \ge b \\ 0, & otherwise \end{cases}$$
(14)

### **Exponential Distribution**

A continuous random variable X is said to have an exponential distribution with parameter  $\lambda > 0$ , shown as  $X \sim \textit{Exponential}(\lambda)$ , if

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x}, & x > 0 \\ 0, & \text{otherwise} \end{cases}, F_X(x) = \begin{cases} 1 - e^{-\lambda x}, & x > 0 \\ 0, & \text{otherwise} \end{cases}$$
(15)

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### Question

### CSIR-UGC NET-June 2013-Problem(72)

Let  $X_1, X_2, ...$  be independent and identically distributed random variables each following a uniform distribution on (0,1). Denote  $T_n = \max(X_1, X_2, ..., X_n)$ . Then, which of the following statements are

true?

- **1**  $T_n$  converges to 1 in probability.
- 2  $n(1-T_n)$  converges in distribution.
- $\sqrt{n}(1-T_n)$  converges to 0 in probability.

#### Solution

Given,  $X_1, X_2, X_3, \ldots$  are independent and identically distributed random variables each following a uniform distribution on (0,1). Let us denote their PDF, CDF by  $f_X(x), F_X(x)$  respectively. Then,

$$f_X(x) = \begin{cases} 1, & 0 < x < 1 \\ 0, & otherwise \end{cases}$$
 (16)

$$F_X(x) = \begin{cases} x, & 0 < x < 1 \\ 1, & x \ge 1 \\ 0, & \text{otherwise} \end{cases}$$
 (17)

### Solution Contd.

As 
$$T_n = \max(X_1, X_2, \dots, X_n) = X_{(n)}$$
, from (13)

$$f_{T_n}(x) = \begin{cases} nx^{n-1}, & 0 < x < 1\\ 0, & otherwise \end{cases}$$
 (18)

Also, from (10)

$$F_{\mathcal{T}_n}(x) = \begin{cases} x^n, & 0 < x < 1 \\ 1, & x \ge 1 \\ 0, & \text{otherwise} \end{cases}$$
 (19)

### Note

Let X be a continuous random variable. If Y = aX + b and a < 0, then

$$F_Y(y) = 1 - F_X\left(\frac{y-b}{a}\right) \tag{20}$$

#### Proof

$$F_Y(y) = \Pr(Y \le y) = \Pr(aX + b \le y)$$
(21)

$$= \Pr\left(X \ge \left(\frac{y-b}{a}\right)\right) (\because a < 0) \tag{22}$$

$$=1-\Pr\left(X\leq \left(\frac{y-b}{a}\right)\right)+\Pr\left(X=\left(\frac{y-b}{a}\right)\right) \tag{23}$$

$$=1-F_X\left(\frac{y-b}{a}\right)+0=1-F_X\left(\frac{y-b}{a}\right) \tag{24}$$

Consider the sequence of random variables  $X_1, X_2, X_3, \ldots$ , such that  $X_n = T_n$ . From (1), we need to evaluate  $\lim_{n \to \infty} \Pr(|T_n - 1| \ge \epsilon), \forall \epsilon > 0$ 

$$\lim_{n \to \infty} \Pr(|T_n - 1| \ge \epsilon) = \lim_{n \to \infty} \Pr(1 - T_n \ge \epsilon)$$
 (25)

$$= \lim_{n \to \infty} \Pr\left(T_n \le 1 - \epsilon\right) = \lim_{n \to \infty} F_{T_n}(1 - \epsilon) \tag{26}$$

$$F_{\mathcal{T}_n}(1-\epsilon) = \begin{cases} (1-\epsilon)^n, & 0 < \epsilon < 1\\ 0, & \epsilon \ge 1 \end{cases}$$
 (27)

$$\lim_{n \to \infty} (1 - \epsilon)^n = 0 \text{ for } 0 < \epsilon < 1$$
 (28)

$$\lim_{n\to\infty} \Pr\left(|T_n - 1| \ge \epsilon\right) = 0, \forall \epsilon > 0$$
 (29)

 $T_n$  converges to 1 in probability.



Consider the sequence of random variables  $X_1, X_2, X_3, \ldots$ , such that  $X_n = n(1 - T_n)$ . From (2), we need to evaluate  $\lim_{n \to \infty} F_{n(1 - T_n)}(x)$ . Substituting a = -n, b = n in (20),

$$F_{n(1-T_n)}(x) = 1 - F_{T_n}\left(1 - \frac{x}{n}\right)$$
 (30)

$$F_{T_n}\left(1 - \frac{x}{n}\right) = \begin{cases} \left(1 - \frac{x}{n}\right)^n, & 0 < x < n \\ 1, & x \le 0 \\ 0, & x \ge n \end{cases}$$
(31)

$$\therefore \lim_{n \to \infty} \left( 1 - \frac{x}{n} \right)^n = e^{-x} \tag{32}$$

# Option 2 Contd.

$$\therefore \lim_{n \to \infty} F_{T_n} \left( 1 - \frac{x}{n} \right) = \begin{cases} e^{-x}, & x > 0 \\ 1, & x \le 0 \end{cases}$$
 (33)

∴ The CDF in (35) represents an exponential distribution with  $\lambda = 1$  ∴  $n(1 - T_n)$  converges in distribution to the random variable  $X \sim Exponential(1)$ .

## Option 2 Contd.

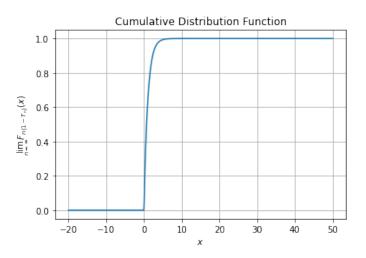


Figure: Plot for the CDF defined in (35)

Consider the sequence of random variables  $X_1, X_2, X_3, \ldots$ , such that  $X_n = n^2(1 - T_n)$ . From (2), we need to evaluate  $\lim_{n \to \infty} F_{n^2(1 - T_n)}(x)$ . Substituting  $a = -n^2$ ,  $b = n^2$  in (20),

$$F_{n^2(1-T_n)}(x) = 1 - F_{T_n}\left(1 - \frac{x}{n^2}\right)$$
 (36)

$$F_{T_n}\left(1 - \frac{x}{n^2}\right) = \begin{cases} \left(1 - \frac{x}{n^2}\right)^n, & 0 < x < n^2 \\ 1, & x \le 0 \\ 0, & x \ge n^2 \end{cases}$$
(37)

$$\therefore \lim_{n \to \infty} \left( 1 - \frac{x}{n^2} \right)^n = 1 \tag{38}$$

# Option 3 Contd.

$$\therefore \lim_{n \to \infty} F_{T_n} \left( 1 - \frac{x}{n^2} \right) = \begin{cases} 1, & x > 0 \\ 1, & x \le 0 \end{cases}$$
 (39)

- ∴ The CDF in (41) is not valid,
- $\therefore n^2(1-T_n)$  does not converge in distribution.

Consider the sequence of random variables  $X_1, X_2, X_3, \ldots$ , such that  $X_n = \sqrt{n}(1-T_n)$ . From (1), we need to evaluate  $\lim_{n \to \infty} \Pr\left(|\sqrt{n}(1-T_n) - 0| \ge \epsilon\right), \forall \epsilon > 0$ 

$$\lim_{n \to \infty} \Pr\left(|\sqrt{n}(1 - T_n) - 0| \ge \epsilon\right) = \lim_{n \to \infty} \Pr\left(1 - T_n \ge \frac{\epsilon}{\sqrt{n}}\right) \tag{42}$$

$$= \lim_{n \to \infty} \Pr\left(T_n \le 1 - \frac{\epsilon}{\sqrt{n}}\right) = \lim_{n \to \infty} F_{T_n}\left(1 - \frac{\epsilon}{\sqrt{n}}\right) \tag{43}$$

$$F_{T_n}\left(1 - \frac{\epsilon}{\sqrt{n}}\right) = \begin{cases} \left(1 - \frac{\epsilon}{\sqrt{n}}\right)^n, & 0 < \epsilon < \sqrt{n} \\ 0, & \epsilon \ge \sqrt{n} \end{cases}$$
(44)

# Option 4 Contd.

$$\lim_{n \to \infty} \left( 1 - \frac{\epsilon}{\sqrt{n}} \right)^n = 0 \text{ for } 0 < \epsilon < \sqrt{n}$$
 (45)

$$\therefore \lim_{n \to \infty} \Pr\left(|\sqrt{n}(1 - T_n) - 0| \ge \epsilon\right) = 0, \forall \epsilon > 0$$
 (46)

 $\therefore \sqrt{n}(1-T_n)$  converges to 0 in probability.

### Solution Contd.

Therefore, options 1), 2), 4) are correct.