#### 1

# Assignment 4

# Suraj - CS20BTECH11050

# Download all python codes from

https://github.com/Suraj11050/Assignments— AI1103/tree/main/Assignment%204/Python %20codes

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# 1 GATE 2021 (ST), Q.17 (STATISTICS SECTION)

If the marginal probability density function of the  $k^{th}$  order statistic of a random sample of size 8 from a uniform distribution on [0, 2] is

$$f(x) = \begin{cases} \frac{7}{32} x^6 (2 - x), & 0 < x < 2, \\ 0, & \text{otherwise,} \end{cases}$$
 (1.0.1)

then *k* equals \_\_\_\_\_

# 2 SOLUTION

**Definition 2.1.** For given statistical sample  $\{X_1, X_2, \dots X_n\}$ , the order statistics is obtained by sorting the sample in ascending order. It denoted as  $\{X_{(1)}, X_{(2)}, \dots X_{(n)}\}$ . The  $k^{th}$  smallest value  $X_{(k)}$  is called  $k^{th}$  order statistic

**Theorem 2.1.** Let  $\{X_1, X_2, \dots X_n\}$  be n i.i.d random variables with common CDF = F(x) and common PDF = f(x), then the marginal probability distribution of  $k^{th}$  order statistic (CDF) is denoted by  $F_{(k,n)}(x)$  and it is given by

$$F_{(k,n)}(x) = \sum_{j=k}^{n} {^{n}C_{j} \times (F(x))^{j} \times (1 - F(x))^{n-j}}$$
(2.0.1)

Proof.

$$F_{(k,n)}(x) = \Pr(X_{(k)} \le x)$$
 (2.0.2)

 $F_{(k,n)}(x) = \Pr(\text{At least k elements have value } \le x)$ (2.0.3)

Since  $Pr(X \le x) = F(x)$ , Let  $Q \sim Bern(F(x))$ 

$$Pr(Q = 1) = F(x)$$
 (2.0.4)

$$\Pr(Q = 0) = 1 - F(x) \tag{2.0.5}$$

Let  $P \sim B(n, F(x))$  taking n trails from Bern(F(x))

$$Pr(P = i) = {}^{n}C_{i} Pr(Q = 1)^{i} Pr(Q = 0)^{n-i}$$
 (2.0.6)

$$\Pr(P = i) = {}^{n}C_{i}F(x)^{i}(1 - F(x))^{n-i}$$
(2.0.7)

Equation (2.0.7) is probability of exactly i R.V of given sample have values  $\leq x$ 

$$F_{(k,n)}(x) = \Pr(P \ge k) = \sum_{j=k}^{n} \Pr(P = j)$$
 (2.0.8)

$$\therefore F_{(k,n)}(x) = \sum_{j=k}^{n} {^{n}C_{j}} (F(x))^{j} (1 - F(x))^{n-j} (2.0.9)$$

**Theorem 2.2.** Let  $\{X_1, X_2, \dots X_n\}$  be n i.i.d random variables with common CDF = F(x) and common PDF = f(x), then the marginal probability density of  $k^{th}$  order statistic (PDF) is denoted by  $f_{(k,n)}(x)$  and it is given by

$$f_{(k,n)}(x) = n^{n-1} C_{k-1} f(x) (F(x))^{k-1} (1 - F(x))^{n-k}$$
(2.0.10)

Proof.

$$\frac{d}{dx}F_{(k,n)}(x) = \frac{d}{dx} \left( \sum_{j=k}^{n} {^{n}C_{j}} \left( 1 - F(x) \right)^{n-j} F(x)^{j} \right)$$
(2.0.11)

$$f_{(k,n)}(x) = \sum_{j=k}^{n} {}^{n}C_{j} (j) (1 - F(x))^{n-j} F(x)^{j-1} f(x)$$
$$- \sum_{j=k}^{n} {}^{n}C_{j} (n-j) (1 - F(x))^{n-j-1} F(x)^{j} f(x)$$
(2.0.12)

$$S_{1} = \sum_{j=k}^{n} \frac{n!}{(n-j)! (j-1)!} (1 - F(x))^{n-j} F(x)^{j-1} f(x) \qquad f_{(k,8)}(x) = \frac{8}{2^{(1+(k-1)+(8-k))}} \times {}^{7}C_{k-1} x^{k-1} (2-x)^{8-k}$$
(2.0.13)

$$S_2 = \sum_{j=k}^{n} \frac{n!}{(n-j-1)! \, j!} (1 - F(x))^{n-j-1} F(x)^j f(x)$$
(2.0.14)

let i = j + 1 change the limits for the summation in equation (2.0.14)

$$S_2 = \sum_{i=k+1}^{n} \frac{n!}{(n-i)! (i-1)!} (1 - F(x))^{n-i} F(x)^{i-1} f(x)$$
(2.0.15)

$$f_{(k,n)}(x) = S_1 - S_2 (2.0.16)$$

$$f_{(k,n)}(x) = \frac{n! f(x) (1 - F(x))^{n-k} F(x)^{k-1}}{(n-k)! (k-1)!}$$
(2.0.17)

$$\therefore f_{(k,n)}(x) = n^{n-1} C_{k-1} (1 - F(x))^{n-k} F(x)^{k-1} f(x)$$
(2.0.18)

# Method 1:

Let  $X \in [0, 2]$  be a random variable of uniform order statistic distribution of sample size 8 then

$$\int_0^2 \Pr(x) \ dx = 1 \tag{2.0.19}$$

$$Pr(x) = \frac{1}{2} \text{ (:: Uniform order)} \quad (2.0.20)$$

The PDF for X is

$$f(x) = \begin{cases} \frac{1}{2}, & 0 < x < 2, \\ 0, & \text{otherwise,} \end{cases}$$
 (2.0.21)

The CDF for X is

$$F(x) = \begin{cases} 0, & x \le 0, \\ \frac{x}{2}, & 0 < x < 2, \\ 1, & x \ge 2 \end{cases}$$
 (2.0.22)

Using theorem (2.2) PDF of  $k^{th}$  order statistic of given sample from equation (2.0.10)

$$f_{(k,n)}(x) = n^{n-1} C_{k-1} \frac{1}{2} \left( \frac{x}{2} \right)^{k-1} \left( 1 - \frac{x}{2} \right)^{n-k}$$
 (2.0.23)

$$f_{(k,8)}(x) = \frac{8}{2^{(1+(k-1)+(8-k))}} \times {}^{7}C_{k-1} x^{k-1} (2-x)^{8-k}$$
(2.0.24)

$$f_{(k,8)}(x) = {}^{7}C_{k-1} \frac{1}{32} x^{k-1} (2-x)^{8-k}$$
 (2.0.25)

Comparing the PDF obtained in equation (2.0.25) with the equation (1.0.1)

$$\frac{1}{32} {}^{7}C_{k-1} (2-x)^{8-k} x^{k-1} = \frac{7}{32} (2-x) x^{6} (2.0.26)$$

$$\therefore k = 7 (2.0.27)$$

Hence the marginal probability density given is  $7^{th}$ order statistic and the value of k is 7

# **Definition 2.2. Uniform order statistics**

Let  $\{X_1, \dots X_n\}$  be i.i.d form a uniform distribution on [0, 1] such that f(x) = 1 and F(x) = x, from theorem (2.2), equation (2.0.10)

$$f_{(k,n)}(x) = n^{n-1} C_{k-1} x^{k-1} (1-x)^{n-k}$$
 (2.0.28)

Since equation (2.0.28) is PDF

$$\int_{0}^{1} n^{n-1} C_{k-1} x^{k-1} (1-x)^{n-k} dx = 1 \qquad (2.0.29)$$

$$\int_{0}^{1} x^{k-1} (1-x)^{n-k} dx = \frac{(k-1)! (n-k)!}{n!} (2.0.30)$$

$$\int_{0}^{1} x^{k-1} (1-x)^{n-k} dx = \frac{\Gamma(k) \Gamma(n-k+1)}{\Gamma((n-k+1)+k)}$$
(2.0.31)

### **Definition 2.3. Beta function**

From definition (2.2), equation (2.0.31) let r = kand s = n - k + 1 The **Beta function** is defined for r, s > 0

$$B(r,s) = \int_{0}^{1} x^{r-1} (1-x)^{s-1} dx = \frac{\Gamma(r)\Gamma(s)}{\Gamma(r+s)}$$
(2.0.32)

# **Beta Distribution**

The Beta distribution is a continuous distribution defined on the range (0, 1) whose PDF given by

$$f(x) = \frac{1}{B(r,s)} x^{r-1} (1-x)^{s-1}$$
 (2.0.33)

where  $\int_{0}^{1} f(x) = 1$  as per definition (2.2)

CDF, Mean value and Variance of Beta distribution

$$F(x) = \frac{\int_{0}^{x} x^{r-1} (1-x)^{s-1}}{B(r,s)} = \frac{B_x(r,s)}{B(r,s)}$$
 (2.0.34)

$$E(x) = \frac{r}{r+s}$$
 (2.0.35)

$$Var(x) = \frac{rs}{(r+s)^2 (r+s+1)}$$
 (2.0.36)

In Uniform order statistics on [0,1] the PDF of  $k^{th}$  order statistic follows Beta distribution with r = k, s = n - k + 1 and PDF is given by

$$f(x) = \frac{1}{B(k, n - k + 1)} x^{k-1} (1 - x)^{(n-k+1)-1}$$
(2.0.37)

#### Method 2:

we know that, PDF of  $k^{th}$  order statistic of a uniform distribution on [0, 1] follows beta distribution

$$\int_{0}^{2} f(x) dx = \int_{0}^{2} \frac{7}{32} x^{6} (2 - x) dx$$
 (2.0.38)

$$\int_{0}^{2} f(x) dx = \int_{0}^{2} 56 \left(\frac{x}{2}\right)^{6} \left(1 - \frac{x}{2}\right) d\left(\frac{x}{2}\right) \quad (2.0.39)$$

Let new random variable be t such that t = x/2, New sample be  $\{T_1, \dots T_8\}$  such that  $T_i = X_i/2$ .

$$f(t) = 56 t^6 (1 - t) (2.0.40)$$

$$\int_{0}^{2} f(x) dx = \int_{0}^{1} f(t) dt = 1$$
 (2.0.41)

The Uniform distribution of new random sample is on [0, 1] such that PDF = 1 and CDF = t f(k, 8)(x) in equation (1.0.1) (after conversion)

$$f_{(k,8)}(t) = \begin{cases} 56 t^6 (1-t), & 0 < t < 1, \\ 0, & \text{otherwise,} \end{cases}$$
 (2.0.42)

Since equation (2.0.42) is a Beta distribution with r = k, s = n - k + 1

$$r - 1 = k - 1 = 6 \tag{2.0.43}$$

$$k = 7$$
 (2.0.44)

Hence the value of k is 7

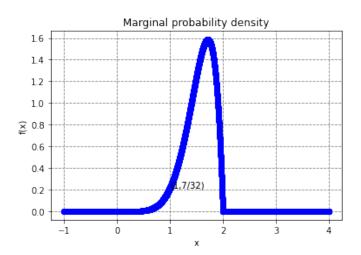


Fig. 1: PDF of  $f_{(7,8)}(x)$ 3.0

2.5

2.0

1.5

1.0

0.5,7/16)

0.0

-1.0

-0.5

0.0

0.5

1.0

1.5

2.0

Fig. 2: PDF of  $f_{(7.8)}(t)$ 

### **Presentation link:**

https:

//github.com/Suraj11050/Assignments-AI1103/tree/main/Assignment4presentation