1

Assignment 4

Suraj - CS20BTECH11050

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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

$$f_{(k,8)}(x) = \begin{cases} \frac{7}{32} x^6 (2-x), & 0 < x < 2, \\ 0, & \text{otherwise,} \end{cases}$$
 (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 = $F_X(x)$ and common PDF = $f_X(x)$, then the marginal probability distribution of k^{th} order statistic (CDF) is denoted by $F_{(k,n)}(x)$ and it is given by

$$F_{(k,n)}(x) = \sum_{j=k}^{n} {}^{n}C_{j} \times (F_{X}(x))^{j} \times (1 - F_{X}(x))^{n-j}$$
(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 $Pr(X \le x) = F_X(x)$, Let $Q \sim Bern(F_X(x))$

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

$$Pr(Q = 0) = 1 - F_X(x)$$
 (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)

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

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

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

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

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

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

Proof.

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

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

$$-\sum_{j=k}^{n} {^{n}C_{j}} (n-j) (1 - F_{X}(x))^{n-j-1} F_{X}(x)^{j} f_{X}(x)$$
(2.0.12)

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

$$S_2 = \sum_{j=k}^{n} \frac{n!}{(n-j-1)! \ j!} (1 - F_X(x))^{n-j-1} F_X(x)^j f_X(x)$$
 Comparing the PDF obtained in equation (2.0.25) with the equation (1.0.1)

let i = j + 1 in equation (2.0.14) (changing limits)

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

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

$$f_{(k,n)}(x) = \frac{n! f_X(x) (1 - F_X(x))^{n-k} F_X(x)^{k-1}}{(n-k)! (k-1)!}$$

$$\therefore f_{(k,n)}(x) = n^{n-1} C_{k-1} (1 - F_X(x))^{n-k} F_X(x)^{k-1} f_X(x)$$
(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

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

The CDF for X is

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

with the equation (1.0.1)

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

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

Hence the marginal probability density given is 7th order statistic and the value of k is 7

Definition 2.2. Beta function

The **Beta function** is defined for $r, s \in \mathbb{R}^+$ by

$$B(r,s) = \int_{0}^{1} x^{r-1} (1-x)^{s-1} dx = \frac{\Gamma(r)\Gamma(s)}{\Gamma(r+s)}$$
(2.0.28)

Definition 2.3. Uniform order statistics

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

$$f_{(k,n)}(x) = n^{n-1} C_{k-1} x^{k-1} (1-x)^{n-k}$$
 (2.0.29)

Lemma 2.3. Since equation (2.0.29) is PDF its definite integral can be given by Beta function as

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

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

$$\int_{0}^{1} x^{k-1} (1-x)^{n-k} dx = \frac{\Gamma(k) \Gamma(n-k+1)}{\Gamma(k+(n-k+1))}$$
(2.0.32)

$$\int_{0}^{1} x^{k-1} (1-x)^{n-k} dx = B(k, n-k+1)$$
 (2.0.33)

Definition 2.4. Beta Distribution

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

$$f(x) = \frac{1}{B(r,s)} x^{r-1} (1-x)^{s-1}$$
 (2.0.34)

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

CDF, Mean value and Variance of Beta distribution

$$F(x) = \frac{\int_{0}^{x} x^{r-1} (1-x)^{s-1}}{B(r,s)} = \frac{B_x(r,s)}{B(r,s)}$$
 (2.0.35)

$$E(x) = \frac{r}{r+s}$$
 (2.0.36)

$$Var(x) = \frac{rs}{(r+s)^2 (r+s+1)}$$
 (2.0.37)

Lemma 2.4. 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

$$f_X(x) = \frac{1}{B(k, n - k + 1)} x^{k-1} (1 - x)^{(n-k+1)-1}$$
(2.0.38)

Method 2:

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

$$\int_{0}^{2} f_X(x) dx = \int_{0}^{2} \frac{7}{32} x^6 (2 - x) dx$$
 (2.0.39)

$$\int_{0}^{2} f_X(x) dx = \int_{0}^{2} 56 \left(\frac{x}{2}\right)^6 \left(1 - \frac{x}{2}\right) d\left(\frac{x}{2}\right) \quad (2.0.40)$$

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

$$f_T(t) = 56 t^6 (1 - t)$$
 (2.0.41)

$$\int_{0}^{2} f_X(x) \, dx = \int_{0}^{1} f_T(t) \, dt = 1 \tag{2.0.42}$$

The Uniform distribution of new random sample is //github.com/Suraj11050/Assignments-AI1103/tree/ on [0, 1] such that PDF = 1 and CDF = tf(k,8)(x) in equation (1.0.1) (after conversion)

$$f_{(k,8)}(t) = \begin{cases} 56 t^6 (1-t), & 0 < t < 1, \\ 0, & \text{otherwise,} \end{cases}$$
 (2.0.43)

Since equation (2.0.43) is a Beta distribution with r = k, s = n - k + 1

$$r - 1 = k - 1 = 6 \tag{2.0.44}$$

$$k = 7$$
 (2.0.45)

Hence the value of k is 7

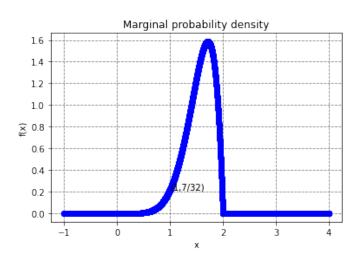


Fig. 1: PDF of $f_{(7.8)}(x)$ 3.0 2.5 2.0 ₽ 1.5 1.0 0.5 5,7/16 0.0 0.0 0.5 1.0 1.5 2.0

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

Presentation link:

https:

main/Assignment4presentation