

# I

## Order Statistics

Def: The order statistics of a random sample  $X_1, \dots, X_n$  are the sample values placed in ascending order, denoted by

$$X_{(1)}, \dots, X_{(n)}$$

Note: Order statistics should be translated as "顺序统计量".

Statistics related to order statistics:

$$1^{\circ} \text{ Sample range } R = X_{(n)} - X_{(1)}$$

$$2^{\circ} \text{ Sample median } M = \begin{cases} X_{(\frac{n+1}{2})}, & n \text{ odd} \\ \frac{1}{2}(X_{(\frac{n}{2})} + X_{(\frac{n}{2}+1)}) & n \text{ even} \end{cases}$$

$$3^{\circ} \text{ Interquartile range } IQR = X_{(n+1-\lceil \frac{n}{4} \rceil)} - X_{(\lceil \frac{n}{4} \rceil)}$$

where  $\{b\}$  is  $b$  round to nearest integer.

Lemma:  $X_1, \dots, X_n$  is a random sample from a continuous distribution  $f(x)$

$$\text{then } (X_1, \dots, X_n) \sim f(x)$$

then the joint density function of  $X_{(1)}, \dots, X_{(n)}$  is

$$f_{X_{(1)}, \dots, X_{(n)}}(x) = \sum_{\pi \in S_n} I(x_{(1)} < \dots < x_{(n)})$$

where  $\pi$  is a permutation of subscripts,  $I(\cdot)$  is the indicator function.

proof: For continuous distribution,

$$P(X_{(1)} < X_{(2)} < \dots < X_{(n)}) =$$

∴ We can regard the sample as distinct.

Let  $I_1, \dots, I_n$  be ordered, nonoverlapping intervals,

$$\begin{aligned}
 & \text{Then } P(X_{(1)} \in I_1, \dots, X_{(m)} \in I_m) \\
 &= P\left(\bigcup_{\pi \in S_n} \{X_{\pi(1)} \in I_1, \dots, X_{\pi(m)} \in I_m\}\right) \\
 &= \sum_{\pi \in S_n} P(X_{\pi(1)} \in I_1, \dots, X_{\pi(m)} \in I_m) \\
 &= \sum_{\pi \in S_n} \int_{I_1} \dots \int_{I_m} f_{\pi}(x) dx \\
 &\quad \text{---} \\
 &\quad \cancel{P(X_{(1)} = x_1, \dots, X_{(m)} = x_m) = p} \\
 &\therefore P(X_{(1)} \leq x_1, \dots, X_{(m)} \leq x_m) = P(X_1 \leq x_{\pi^{-1}(1)}, \dots, X_n \leq x_{\pi^{-1}(m)})
 \end{aligned}$$

Define  $F_{\pi}(x) = F(x_{\pi^{-1}(1)}, \dots, x_{\pi^{-1}(m)})$ , and

$$\cancel{X_{\pi}} = (X_{\pi(1)}, \dots, X_{\pi(m)})$$

then  $X_{\pi} \sim F_{\pi}$

If density exists, then  $f_{\pi}(x) = f(x_{\pi^{-1}})$

$$\begin{aligned}
 \therefore \text{LHS} &= \sum_{\pi \in S_n} \int_{I_1} \dots \int_{I_m} f_{\pi}(x_{\pi^{-1}}) dx \\
 &= \int_{I_1} \dots \int_{I_m} \sum_{\pi \in S_n} f(x_{\pi^{-1}}) dx
 \end{aligned}$$

The joint density function of order statistics is

$$\begin{aligned}
 \cancel{P(X_{(1)}, \dots, X_{(m)})} &= \sum_{\pi \in S_n} P(x_{\pi^{-1}}) I(x_1 < \dots < x_m) \\
 f_{X_{(1)}, \dots, X_{(m)}}(x) &= \sum_{\pi \in S_n} f(x_{\pi^{-1}}) I(x_1 < \dots < x_m)
 \end{aligned}$$

□.

Note:  $\{\underline{x}: x_1 < \dots < x_n\}$  is a hyper-tetrahedron on which the order statistics are defined.

~~Special cases of pdf of order statistics.~~

~~$x$  is exchangeable.~~  $f(x) = \frac{f(x)}{n!} I(x_1 < \dots < x_n)$

Thm:  $\cancel{X_1, \dots, X_n}$  Random Sample  $n! \prod_{i=1}^n f(x_i) I(x_1 < \dots < x_n)$

Note:  $\cancel{X_1, \dots, X_n}$  Random Sample from Uniform(0,1):  $n! \prod_{i=1}^n I(x_i < \dots < x_n \& i < n)$

Thm:  $X_1, \dots, X_n$  is a random sample from a continuous distribution with cdf  $F(x)$ , and pdf  $f(x)$ . Then the pdf of  $X_{(r)}$  is

$$f_{X_{(r)}}(x) = \binom{n}{r-1, n-r} F(x)^{r-1} f(x) (1-F(x))^{n-r}$$

The cdf of  $X_{(r)}$  is

$$F_{X_{(r)}}(x) = \sum_{k=r}^n \binom{n}{k} F(x)^k (1-F(x))^{n-k}$$

proof:

$$\begin{aligned} f_{X_{(r)}}(x) &= \int_{\mathbb{R}^{n-r}} n! \prod_{i=1}^{r-1} f(x_i) \cdot f(x) \cdot \prod_{j=r+1}^n f(x_j) \\ &\quad I(x_1 < \dots < x_{r-1} < x < x_r < \dots < x_n) dx_1 \dots dx_{r-1} dx_r dx_{r+1} \dots dx_n \\ &= n! f(x) \cdot \left[ \int_{\mathbb{R}^{n-r}} \prod_{i=1}^{r-1} f(x_i) I(x_1 < \dots < x_{r-1} < x) dx_1 \dots dx_{r-1} \right] \\ &\quad \cdot \left[ \int_{\mathbb{R}^{n-r}} \prod_{j=r+1}^n f(x_j) I(x < x_{r+1} < \dots < x_n) dx_{r+1} \dots dx_n \right] \\ &= n! f(x) \cdot \left[ \int_{-\infty}^x dx_{r-1} \int_{-\infty}^{x_{r-1}} dx_{r-2} \dots \int_{-\infty}^{x_2} dx_1 \cdot \prod_{i=1}^{r-1} f(x_i) \right] \\ &\quad \cdot \left[ \int_x^{+\infty} dx_{r+1} \int_{x_{r+1}}^{+\infty} dx_{r+2} \dots \int_{x_{n-1}}^{+\infty} dx_n \cdot \prod_{j=r+1}^n f(x_j) \right] \\ &= n! f(x) \frac{F(x)^{r-1}}{(r-1)!} \frac{(1-F(x))^{n-r}}{(n-r)!} \\ &= \binom{n}{r-1, n-r} F(x)^{r-1} f(x) (1-F(x))^{n-r} \end{aligned}$$

where  $\binom{n}{r-1, n-r} = \frac{n!}{(r-1)! \cdot 1! \cdot (n-r)!}$

$$\begin{aligned} F_{X_{(r)}}(x) &= P(X_{(r)} \leq x) \\ &= P\left(\bigcup_{i=r}^n \{i : X_i \leq x\} = j\right) \\ &= \sum_{j=r}^n P(\{i : X_i \leq x\} = j) \end{aligned}$$

$$= \sum_{j=r}^n \binom{n}{j} F(x)^j (1-F(x))^{n-j}$$

□.

Thm: ~~If~~  $X_1, \dots, X_n$  is a random sample from a continuous distribution with cdf  $F(x)$  and pdf  $f(x)$ . Then the joint pdf of  $X_{(i)}$  and  $X_{(j)}$  ( $1 \leq i < j \leq n$ ) is

$$f_{X_{(i)}, X_{(j)}}(u, v) = \binom{n}{i-1, 1, j-i-1, 1, n-j} F(u)^{i-1} f(u) (F(v) - F(u))^{j-i-1} f(v) (1-F(v))^{n-j} I(u < v)$$

Thm: ~~The pdf of uniform order statistic  $U_m$  is~~

Uniform order statistic

$$U_{(r)} \sim B(r, n-r+1) = \frac{\Gamma(n+1)}{\Gamma(r)\Gamma(n-r+1)} u^{r-1} (1-u)^{n-r}$$

proof:  $f_{U_{(r)}}(u) = \binom{n}{r-1, n-r} F_U(u)^{r-1} f_U(u) (1-F_U(u))^{n-r}$

$$= \frac{n!}{(r-1)!(n-r)!} u^{r-1} (1-u)^{n-r}$$

$$= \frac{\Gamma(n+1)}{\Gamma(r)\Gamma(n-r+1)} u^{r-1} (1-u)^{n-r}$$

□.