

Learning Objectives

- ▶ Distribution of $X_{(1)}$
- ▶ Distribution of $X_{(n)}$
- ▶ Distribution of $(X_{(1)}, \dots, X_{(n)})$
- ▶ Distribution of $X_{(r)}$
- ▶ Joint Distribution of $X_{(r)}$ and $X_{(s)}$
- ▶ Distribution of sample range $R = X_{(n)} - X_{(1)}$

Order Statistics (O.S.)

- ▶ X_1, X_2, \dots, X_n : random sample of size n
- ▶ F_X : cdf (continuous) of the population
- ▶ $X_{(1)}, X_{(2)}, \dots, X_{(n)}$: corresponding order statistics such that

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

- ▶ $X_{(r)}$: r^{th} order statistic, $1 \leq r \leq n$.

Distribution of $X_{(1)} = \min\{X_1, \dots, X_n\}$

Let $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} F$. Then, the distribution of $X_{(1)}$ is

$$f_{X_{(1)}}(y_1) = n[1 - F(y_1)]^{n-1}f(y_1).$$

Proof:

$$\begin{aligned} P(X_{(1)} > y_1) &= P(X_1 > y_1, \dots, X_n > y_n) = \prod_{i=1}^n P(X_i > y_i) \\ &= [1 - F(y_1)]^n \end{aligned}$$

So, $F_{X_{(1)}}(y_1) = 1 - [1 - F(y_1)]^n$.

Since F is absolutely continuous, we have the pdf of $X_{(1)}$ as

$$f_{X_{(1)}}(y_1) = n[1 - F(y_1)]^{n-1}f(y_1).$$

Distribution of $X_{(n)} = \max\{X_1, \dots, X_n\}$

Let $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} F$. Then, the distribution of $X_{(n)}$ is

$$f_{X_{(n)}}(y_n) = n[F(y_n)]^{n-1}f(y_n).$$

Proof:

$$\begin{aligned} P(X_{(n)} \leq y_n) &= P(X_1 \leq y_1, \dots, X_n \leq y_n) = \prod_{i=1}^n P(X_i \leq y_i) \\ &= [F(y_n)]^n \end{aligned}$$

So, $F_{X_{(n)}}(y_n) = [F(y_n)]^n$.

Since F is absolutely continuous, we have the pdf of $X_{(n)}$ as

$$f_{X_{(n)}}(y_n) = n[F(y_n)]^{n-1}f(y_n).$$

Examples

- ▶ Let $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} \text{Exp}(\lambda)$. Then, $X_{(1)} \sim \text{Exp}(n\lambda)$.
- ▶ Let $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} U[0, \theta]$. Then, the distribution of $X_{(n)}$ is

$$f_{X_{(n)}}(y_n) = \frac{ny_n^n}{\theta^n}, \quad 0 \leq y_n \leq 1.$$

Joint distribution of $X_{(1)}, \dots, X_{(n)}$

Let X_1, \dots, X_n is a random sample from continuous distribution F with pdf f . Define

$$Y_i = X_{(i)}, \quad i = 1, \dots, n.$$

Then, the joint distribution of (Y_1, \dots, Y_n) is

$$f_Y(\mathbf{y}) = n! \prod_{i=1}^n f(y_i), \quad -\infty < y_1 < \dots < y_n < \infty.$$

Proof: We know that the joint pdf of $\mathbf{X} = (X_1, \dots, X_n)$ is

$$f_{\mathbf{X}}(\mathbf{x}) = \prod_{i=1}^n f(x_i)$$

Joint distribution of $X_{(1)}, \dots, X_{(n)}$ contd.

There are $n!$ inverse images of the above transformation, such as,

$$A_1 = \{\mathbf{x} : x_1 < x_2 < x_3 < \dots < x_n\}$$

$$A_2 = \{\mathbf{x} : x_2 < x_1 < x_3 < \dots < x_n\}$$

$$\vdots \qquad \qquad \qquad \vdots$$

$$A_n = \{\mathbf{x} : x_n < x_{n-1} < \dots < x_1\}$$

In the first region A_1 , the inverse image is

$$x_1 = y_1$$

$$x_2 = y_2$$

$$\vdots \quad \vdots$$

$$x_n = y_n$$

and the corresponding Jacobian $J = 1$.

Joint distribution of $X_{(1)}, \dots, X_{(n)}$ contd.

In the second region A_2 , the inverse image is

$$x_1 = y_2$$

$$x_2 = y_1$$

$$\vdots \quad \vdots$$

$$x_n = y_n$$

and the corresponding Jacobian $J = -1$.

Similarly, in the last region A_n , the inverse image is

$$x_1 = y_n$$

$$x_2 = y_{n-1}$$

$$\vdots \quad \vdots$$

$$x_n = y_1$$

and the corresponding Jacobian $J = (-1)^n$.

Joint distribution of $X_{(1)}, \dots, X_{(n)}$ contd.

Thus, in all the cases, $|J| = 1$.

Further, the density in every region is $\prod_{i=1}^n f(y_i)$.

Since there are $n!$ region, the pdf of (Y_1, \dots, Y_n) is

$$f_Y(y) = n! \prod_{i=1}^n f(y_i), \quad -\infty < y_1 < \dots < y_n < \infty.$$

Distribution of $X_{(r)}$, $1 \leq r \leq n$

Let $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} F$. Then, the distribution of $Y_r = X_{(r)}$ is

$$f_{Y_r}(y_r) = \frac{n!}{(r-1)!(n-r)!} [F(y_r)]^{r-1} [1 - F(y_r)]^{n-r} f(y_r),$$

where $-\infty < y_r < \infty$.

Proof: We know that the joint distribution of (Y_1, \dots, Y_n) is

$$f_Y(\mathbf{y}) = n! \prod_{i=1}^n f(y_i), \quad -\infty < y_1 < \dots < y_n < \infty.$$

In order to find the marginal pdf of Y_r , $f_Y(\mathbf{y})$ has to be integrated w.r.t. $(y_1, \dots, y_{r-1}, y_{r+1}, \dots, y_n)$, i.e.,

Distribution of $X_{(r)}$ contd.

$$f_{Y_r}(y_r) = \int_{y_r}^{\infty} \cdots \int_{y_{n-2}}^{\infty} \int_{y_{n-1}}^{\infty} \int_{-\infty}^{y_r} \cdots \int_{-\infty}^{y_2} n! \prod_{i=1}^n f(y_i) dy_1 \cdots dy_{r-1} dy_n dy_{n-1} \cdots dy_{r+1}$$

Now, integrating

$$f(y_1) \text{ w.r.t. } y_1 \longrightarrow F(y_2)$$

$$F(y_2)f(y_2) \text{ w.r.t. } y_2 \longrightarrow \frac{1}{2}[F(y_3)]^2$$

$$\frac{1}{2}F^2(y_3)f(y_3) \text{ w.r.t. } y_3 \longrightarrow \frac{1}{3!}[F(y_4)]^3$$

\vdots

$$??? \text{ w.r.t. } y_{r-1} \longrightarrow \frac{1}{(r-1)!}[F(y_r)]^{r-1}.$$

Distribution of $X_{(r)}$ contd.

Further, integrating

$$\begin{aligned} f(y_n) \text{ w.r.t. } y_n &\longrightarrow [1 - F(y_{n-1})] \\ [1 - F(y_{n-1})]f(y_{n-1}) \text{ w.r.t. } y_{n-1} &\longrightarrow \frac{1}{2}[1 - F(y_{n-2})]^2 \\ \frac{1}{2}[1 - F(y_{n-1})]^2 f(y_{n-2}) \text{ w.r.t. } y_{n-2} &\longrightarrow \frac{1}{3!}[1 - F(y_{n-3})]^3 \\ &\vdots \\ ??? \text{ w.r.t. } y_{r+1} &\longrightarrow \frac{1}{(n-r)!}[1 - F(y_r)]^{n-r}. \end{aligned}$$

Thus combining the above terms, we get:

$$f_{Y_r}(y_r) = \frac{n!}{(r-1)!(n-r)!} [F(y_r)]^{r-1} [1 - F(y_r)]^{n-r} f(y_r),$$

$$-\infty < y_r < \infty.$$

Remark

- Note that

$$\begin{aligned} f_{Y_r}(y_r) &= \frac{n!}{(r-1)!(n-r)!} [F(y_r)]^{r-1} [1 - F(y_r)]^{n-r} f(y_r) \\ &= \frac{1}{\text{Be}(r, n-r+1)} [F(y_r)]^{r-1} [1 - F(y_r)]^{n-r} f(y_r), -\infty < y_r < \infty. \end{aligned}$$

- **Special case:** $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} U[0, 1]$.

$$f_{Y_r}(y_r) = \frac{1}{\text{Be}(r, n-r+1)} y_r^{r-1} (1 - y_r)^{n-r}, \quad 0 < y_r < 1.$$

Thus, $Y_r \sim \text{Be}(r, n-r+1)$.

Joint distribution of $X_{(r)}$ and $X_{(s)}$

Let $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} F$. Then, the distribution of

$$(Y_r, Y_s) = (X_{(r)}, X_{(s)}), \quad r < s,$$

is

$$f_{Y_r, Y_s}(y_r, y_s) = \frac{n!}{(r-1)!(n-s)!(s-r-1)!} \\ [F(y_r)]^{r-1} [1 - F(y_s)]^{n-r} [F(y_s) - F(y_r)]^{s-r-1} f(y_r) f(y_s),$$

where $-\infty < y_r < y_s < \infty$.

Distribution of sample range $R = X_{(n)} - X_{(1)}$

Let $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} F$. Then, the desired pdf of $R = X_{(n)} - X_{(1)}$ is

$$f_R(r) = n(n-1) \int_r^\infty [F(s) - F(s-r)]^{n-2} f(s-r) f(s) ds.$$

Proof: With $r = 1$ and $s = n$, the joint pdf of $Y_1 = X_{(1)}$ and $Y_n = X_{(n)}$ is

$$\begin{aligned} f_{Y_1, Y_n}(y_1, y_n) \\ = n(n-1)[F(y_n) - F(y_1)]^{n-2} f(y_1) f(y_n), \quad -\infty < y_1 < y_n < \infty \end{aligned}$$

Distribution of sample range R contd.

Now, define the transformation $R = Y_n - Y_1$ and $S = Y_n$.
The inverse transformation is $Y_1 = S - R$ and $Y_n = S$, and
the corresponding Jacobian is $|J| = 1$.

Thus, the joint pdf of R and S is

$$f_{R,S}(r,s) = n(n-1)[F(s) - F(s-r)]^{n-2}f(s-r)f(s), \quad 0 < r < s$$

The desired pdf of R is

$$f_R(r) = n(n-1) \int_r^{\infty} [F(s) - F(s-r)]^{n-2} f(s-r) f(s) ds$$

Distribution of sample range R contd.

- **Special case:** $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} U[0, 1]$.

The pdf of R is

$$f_R(r) = n(n-1) \int_r^1 r^{n-2} ds = n(n-1)r^{n-2}(1-r), \quad 0 < r < 1$$

Thus, $R \sim Be(n-1, 2)$.