

Order Statistics (Section 5.4)

Let X_1, X_2, \dots, X_n be any values.
The order statistics $X_{(1)}, X_{(2)}, \dots, X_{(n)}$
are the values arranged in ascending
order.

Theorem

If X_1, \dots, X_n are iid with cdf F ,
then $F_{X_{(j)}}(x) = \sum_{k=j}^n \binom{n}{k} (F(x))^k (1-F(x))^{n-k}$

Proof:

$$P(X_{(j)} \leq x) = P(\# \text{ of } X_i \leq x \text{ is at least } j)$$

$$= P(Y \geq j)$$

$$\text{where } Y = \sum_{i=1}^n I(X_i \leq x)$$

$$\sim \text{Binomial}(n, p = F(x))$$

Note: Special Case $F_{X_{(n)}}(x) = (F(x))^n$

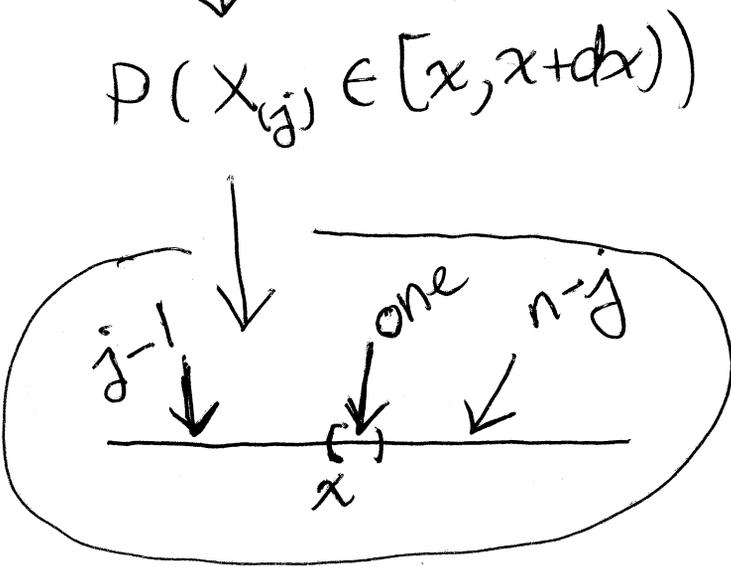
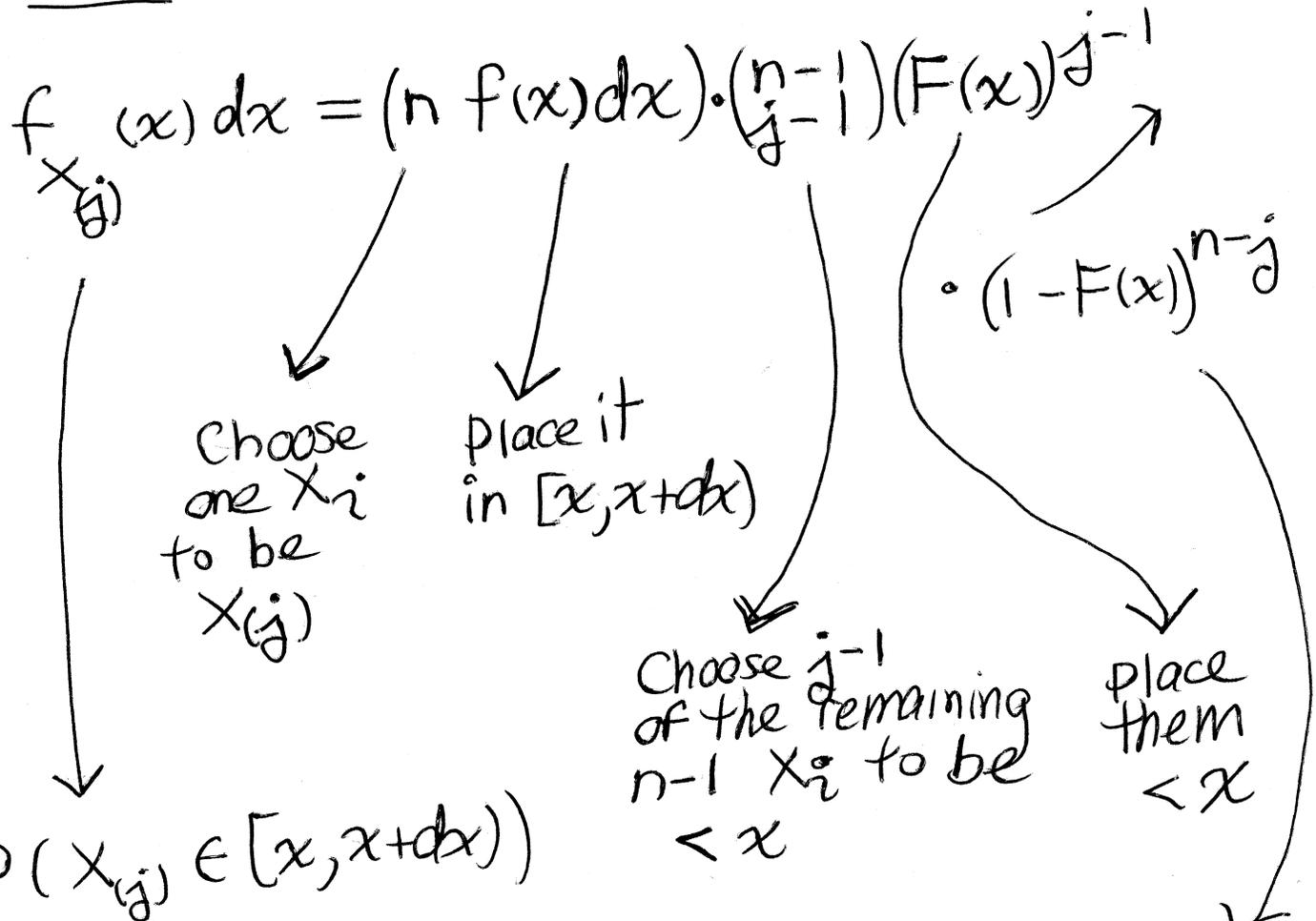
we already knew.

Corollary

If X_1, \dots, X_n are iid with pdf f ,

then $f_{X_{(j)}}(x) = n \binom{n-1}{j-1} f(x) (F(x))^{j-1} (1-F(x))^{n-j}$.

"Proof":



The Real Proof:

$$f_{X_{(j)}}(x) = \frac{d}{dx} F_{X_{(j)}}(x)$$

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

$$\left\{ k F(x)^{k-1} (1-F(x))^{n-k} f(x) \right.$$

$$\left. - (n-k) F(x)^k (1-F(x))^{n-k-1} f(x) \right\}$$

$$\binom{n}{k} k = \binom{n-1}{k-1} n \quad \text{for } k=1, \dots, n$$

($k > 0$)

$$\binom{n}{k} (n-k) = \binom{n-1}{k} n \quad \text{for } k=0, \dots, n-1$$

($k < n$)

(True always if $\binom{a}{b} \equiv 0$
for $b < 0$ or $b > a$.)

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

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

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

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

$$= \underbrace{\left(\text{Term for } l=j-1 \right)}_{=0 \text{ since } \binom{n-1}{n} = 0} - \underbrace{\left(\text{Term for } k=n \right)}_{=0 \text{ since } \binom{n-1}{n} = 0}$$

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

as desired

Special Case:

When F is Uniform(0,1)

$$F(x) = x \text{ for } 0 < x < 1$$

$$f(x) = 1$$

pdf

$$f_{X_{(j)}}(x) = n \binom{n-1}{j-1} 1 \cdot x^{j-1} (1-x)^{n-j}$$

$$= \frac{\Gamma(n+1)}{\Gamma(j) \Gamma(n-j+1)} x^{j-1} (1-x)^{(n-j+1)-1} \text{Beta}(j, n-j+1)$$

Note: (Connection with HW exercise 2.40)

$$F_{X_{(j)}}(x) = \int_0^x f_{X_{(j)}}(u) du$$

$$\begin{aligned} \sum_{k=0}^n \binom{n}{k} F(x)^k (1-F(x))^{n-k} \\ = \int_0^x n \binom{n-1}{j-1} f(u) F(u)^{j-1} (1-F(u))^{n-j} du \end{aligned}$$

For Uniform (0,1) this becomes

$$\begin{aligned} \sum_{k=0}^n \binom{n}{k} x^k (1-x)^{n-k} \\ = \int_0^x n \binom{n-1}{j-1} u^{j-1} (1-u)^{n-j} du \end{aligned}$$

which is a close relative of the result in Exercise 2.40.

Connection with [2.40]



$$P(X_{(j)} > x) = 1 - P(X_{(j)} < x)$$

$$\sum_{k=0}^{j-1} \binom{n}{k} x^k (1-x)^{n-k}$$

$$= \int_x^1 n \binom{n-1}{j-1} u^{j-1} (1-u)^{n-j} du$$

$$= \int_0^{1-x} n \binom{n-1}{j-1} (1-z)^{j-1} z^{n-j} dz$$
$$= \binom{n}{j-1} (1-x)^{n-(j-1)}$$

Multivariate Transformations

(p. 185)

$$X = (X_1, X_2, \dots, X_p), \text{ pdf } f_X(x)$$

$$U = (U_1, U_2, \dots, U_p)$$

$$U = g(X) \text{ where } g: \mathbb{R}^p \rightarrow \mathbb{R}^p$$

$$\mathcal{A} = \{x: f_X(x) > 0\}$$

$$\mathcal{B} = \{u: g(x) = u \text{ for some } x \in \mathcal{A}\}$$

If $g: \mathcal{A} \rightarrow \mathcal{B}$ is 1-1 and "smooth"
with $h = g^{-1}$, then

$$f_U(u) = f_X(h(u)) |J_h(u)| \text{ for } u \in \mathcal{B}.$$

If $\mathcal{A} = A_0 \cup A_1 \cup \dots \cup A_k$ disjoint

with $P(X \in A_0) = 0$

and $g_i: A_i \rightarrow \mathcal{B}$ is 1-1, onto, "smooth"

(where g_i is g restricted to A_i)

with inverse $h_i = g_i^{-1}$ where $h_i: \mathcal{B} \rightarrow A_i$,

then

$$f_U(u) = \sum_{i=1}^k f_X(h_i(u)) |J_{h_i}(u)| \text{ for } u \in \mathcal{B}.$$

Order statistics

Sample size $n=2$

$X = (X_1, X_2)$; X_1, X_2 iid with pdf f

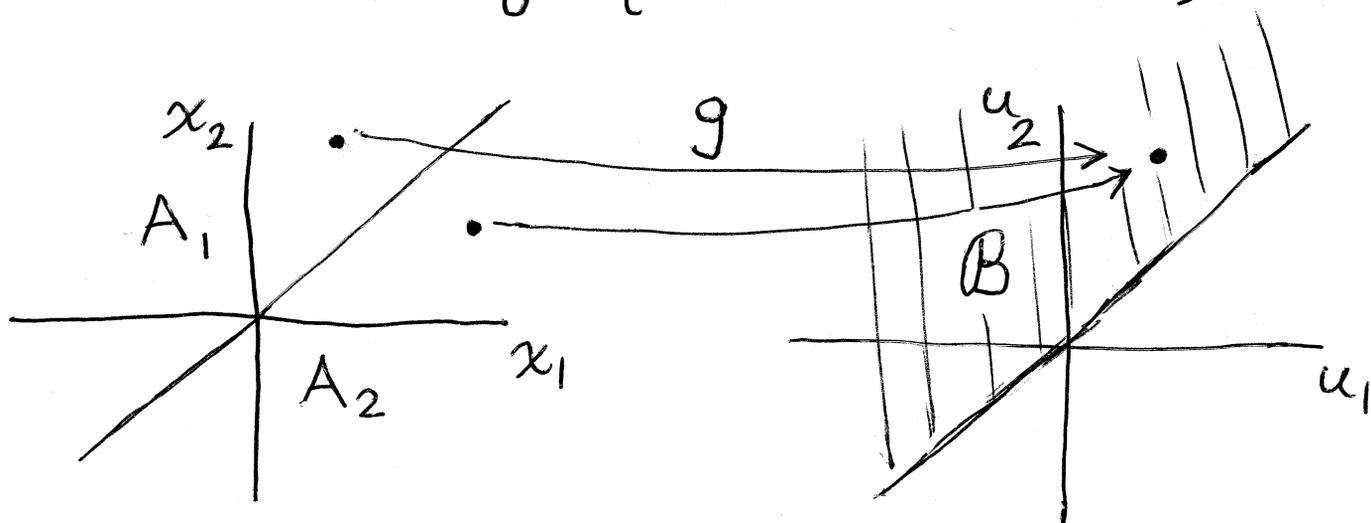
$$\begin{aligned} \text{so that } f_X(x) &= f_{X_1, X_2}(x_1, x_2) \\ &= f(x_1)f(x_2) \end{aligned}$$

$$U = (U_1, U_2) = (X_{(1)}, X_{(2)}) = g(X)$$

where

$$\begin{aligned} g(x) &= g(x_1, x_2) \begin{array}{l} \nearrow g_1 \rightarrow A_1 \\ \searrow g_2 \rightarrow A_2 \end{array} \\ &= \begin{cases} (x_1, x_2) & \text{if } x_1 < x_2 \\ (x_2, x_1) & \text{if } x_2 < x_1 \end{cases} \end{aligned}$$

$$\text{and } A_0 = \{(x_1, x_2) : x_1 = x_2\}.$$



Thus $B = \{(u_1, u_2) : u_1 < u_2\}$

$$h_1: B \rightarrow A_1, \quad h_1 = g_1^{-1}$$

$$h_1(u_1, u_2) = (u_1, u_2)$$

$$h_2: B \rightarrow A_2, \quad h_2 = g_2^{-1}$$

$$h_2(u_1, u_2) = (u_2, u_1).$$

It is easily seen that

$$|J_{h_1}(u)| = |J_{h_2}(u)| = 1.$$

For example

$$x = h_2(u) \Rightarrow \begin{matrix} x_1 = u_2 \\ x_2 = u_1 \end{matrix} \Rightarrow J_{h_2}(u) =$$

$$\begin{vmatrix} \frac{\partial x_1}{\partial u_1} & \frac{\partial x_1}{\partial u_2} \\ \frac{\partial x_2}{\partial u_1} & \frac{\partial x_2}{\partial u_2} \end{vmatrix} = \begin{vmatrix} 0 & 1 \\ 1 & 0 \end{vmatrix} = -1.$$

$$\text{Also } f_x(h_1(u)) = f_x(h_2(u)) = f(u_1)f(u_2).$$

$$\begin{aligned} \text{For example, } f_x(h_2(u)) &= f_{x_1, x_2}(u_2, u_1) \\ &= f(u_2)f(u_1) = f(u_1)f(u_2). \end{aligned}$$

Thus

$$\begin{aligned}f_U(u) &= \sum_{i=1}^2 f_X(h_i(u)) |J_{h_i}(u)| \text{ for } u \in \mathcal{B} \\ &= 2f(u_1)f(u_2) \text{ for } u_1 < u_2 \\ &\quad \text{(zero otherwise)} \\ &= 2f(u_1)f(u_2) I(u_1 < u_2)\end{aligned}$$

Sample size $n=3$

$X = (X_1, X_2, X_3)$ where X_1, X_2, X_3 are iid with pdf f so that

$$f_X(x) = f_{X_1, X_2, X_3}(x_1, x_2, x_3) = f(x_1)f(x_2)f(x_3)$$

$$\begin{aligned}U &= (U_1, U_2, U_3) = (X_{(1)}, X_{(2)}, X_{(3)}) \\ &= g(X) \text{ where } g(x) = g(x_1, x_2, x_3)\end{aligned}$$

$$= \begin{cases} (x_1, x_2, x_3) & \text{if } x_1 < x_2 < x_3 \\ (x_2, x_1, x_3) & \text{if } x_2 < x_1 < x_3 \\ (x_1, x_3, x_2) & \text{if } x_1 < x_3 < x_2 \\ (x_3, x_2, x_1) & \text{if } x_3 < x_2 < x_1 \\ (x_2, x_3, x_1) & \text{if } x_2 < x_3 < x_1 \\ (x_3, x_1, x_2) & \text{if } x_3 < x_1 < x_2 \end{cases}$$