

5.23

$$\begin{aligned}
 P(Z > z) &= \sum_{x=1}^{\infty} P(Z > z|x)P(X = x) \\
 &= P(U_1 > z, \dots, U_x > z|x)P(X = x) \\
 &= \sum_{x=1}^{\infty} \prod_{i=1}^x P(U_i > z)P(X = x) \quad \text{by independence of the } U_i\text{'s} \\
 &= \sum_{x=1}^{\infty} P(U_i > z)^x P(X = x) \\
 &= \sum_{x=1}^{\infty} (1-z)^x \frac{1}{(e-1)x!} \\
 &= \frac{1}{(e-1)} \sum_{x=1}^{\infty} \frac{(1-z)^x}{x!} \\
 &= \frac{e^{1-z} - 1}{e-1}, \quad 0 < z < 1
 \end{aligned}$$

5.24

Use  $f_X(x) = 1/\theta, F_X(x) = x/\theta, 0 < x < 1$ .

Let  $Y = X_{(n)}, Z = X_{(1)}$ . Then, from Theorem 5.4.6,

$$f_{Z,Y}(z, y) = \frac{n!}{0!(n-2)!0!} \frac{1}{\theta} \frac{1}{\theta} \left(\frac{z}{\theta}\right)^0 \left(\frac{y-z}{\theta}\right)^{n-2} \left(1 - \frac{y}{\theta}\right)^0 = \frac{n(n-1)}{\theta^n} (y-z)^{n-2}, \quad 0 < z < y < \theta.$$

Now let  $W = Z/Y, Q = Y$ . Then  $Y = Q, Z = WQ$ , and  $|J| = q$ . Therefore,

$$f_{W,Q}(w, q) = \frac{n(n-1)}{\theta^n} (q - qw)^{n-2} q = \frac{n(n-1)}{\theta^n} (1-w)^{n-2} q^{n-1}, \quad 0 < w < 1, 0 < q < \theta.$$

The joint pdf factors into functions of w and q, hence W and Q are independent.

5.27

- a.  $f_{X(i)|X(j)}(u|v) = \frac{f_{X(i),X(j)}(u,v)}{f_{X(j)}(v)}$ . Consider two cases, depending on which of i or j is greater. Using the formulas from Theorem 5.4.4 and 5.4.6, and after cancellation, we obtain the following.

(i) If  $i < j$ ,

$$\begin{aligned} & f_{X^{(i)}|X^{(j)}}(u|v) \\ &= \frac{(j-1)!}{(i-1)!(j-1-i)!} f_X(u) F_X^{i-1}(u) (F_X(v) - F_X(u))^{j-i-1} F_X^{1-j}(v) \\ &= \frac{(j-1)!}{(i-1)!(j-1-i)!} \frac{f_X(u)}{F_X(v)} \left( \frac{F_X(u)}{F_X(v)} \right)^{i-1} \left( 1 - \frac{F_X(u)}{F_X(v)} \right)^{j-i-1}, u < v. \end{aligned}$$

Note that this is the pdf of the  $i$ th order statistic from a sample of size  $j-1$ , from a population with pdf given by the truncated distribution,  $f(u) = \frac{f_X(u)}{F_X(v)}$ ,  $u < v$ .

(ii) If  $i > j$  and  $u > v$ ,

$$\begin{aligned} & f_{X^{(i)}|X^{(j)}}(u|v) \\ &= \frac{(n-j)!}{(n-1)!(i-1-j)!} f_X(u) (1 - F_X(u))^{n-i} (F_X(u) - F_X(v))^{i-1-j} (1 - F_X(v))^{j-n} \\ &= \frac{(n-j)!}{(i-j-1)!(n-i)!} \frac{f_X(u)}{1 - F_X(v)} \left( \frac{F_X(u) - F_X(v)}{1 - F_X(v)} \right)^{i-j-1} \left( 1 - \frac{F_X(u) - F_X(v)}{1 - F_X(v)} \right)^{n-i}. \end{aligned}$$

This is the pdf of the  $(i-j)$ th order statistic from a sample of size  $n-j$ , from a population with pdf given by the truncated distribution,  $f(u) = \frac{f_X(u)}{1 - F_X(v)}$ ,  $u < v$ .

## Additional Prob

### 1

Let  $F(X_{(n)}) = U_{(n)} \sim \text{unif}(0, 1)$ ,  $F(X_{(1)}) = U_{(1)} \sim \text{unif}(0, 1)$ , then the joint pdf of  $U_{(n)}$  and  $U_{(1)}$  as  $f_{(1),(n)}(w, z)$  is follow:

$$\begin{aligned} f_{(1),(n)}(w, z) &= \frac{n!}{(n-2)!} f(w) (F(z) - F(w))^{n-2} f(z) \\ &= n(n-1)(z-w)^{n-2}, \quad 0 < w < z < 1. \\ P((X_{(n)}) - (X_{(1)}) \geq p) &= \int_0^{1-p} \int_{w+p}^1 n(n-1)(z-w)^{n-2} dz dw \\ &= \int_0^{1-p} n(n-1) \frac{(z-w)^{n-1}}{n-1} \Big|_{w+p}^1 dw \\ &= \int_0^{1-p} n((1-w)^{n-1} - p^{n-1}) dw \\ &= -n \cdot \frac{(1-w)^n}{n} \Big|_0^{1-p} - np^{n-1}(1-p) \\ &= 1 - p^n - np^{n-1}(1-p) \end{aligned}$$

a.

$$\begin{aligned}
f(x) &= k\Phi(x)\sin^2(x) \\
h(x) &= \Phi(x)\sin^2(x) \\
g(x) &= \Phi(x) \\
\frac{h(x)}{g(x)} &= \sin^2(x) \leq 1 \Rightarrow M = 1.
\end{aligned}$$

Then to generate  $X$  from  $kh(x)$ , conduct the following steps:

- 1) generate  $X^*$  from  $g$ .
- 2) generate  $U \sim \text{unif}(0, 1)$ .
- 3) if  $Ug(X^*) < h(X^*)$ , then  $X = X^*$ , else return to 1).

b. To approximate  $EX^2$ , we could do the following:

- 1) use algorithm in a) to generate a large size iid sample  $\{X_i\}_{i=1}^n$ .
- 2) compute

$$\frac{1}{n} \sum_{i=1}^n X_i^2 \cdot \frac{h(X_i)}{g(X_i)} = \frac{1}{n} \sum_{i=1}^n X_i^2 \sin^2(X_i) \quad (1)$$

$$\frac{1}{n} \sum_{i=1}^n \frac{h(X_i)}{g(X_i)} = \frac{1}{n} \sum_{i=1}^n \sin^2(X_i) \quad (2)$$

Then (1)/(2)  $\rightarrow EX^2 = \text{Var}X$  in probability.

c. (First method)

Use empirical cdf to approximate  $P(0.3 < X < 1.2)$ , that is,  $P(0.3 < X < 1.2) = \frac{1}{n} \sum_{i=1}^n I(0.3 < X_i < 1.2)$ , where  $\{X_i\}_{i=1}^n$  are iid random variables.

(Second method)

Use importance sampling method to do the following:

- 1) use algorithm in a) to generate a large size random sample  $\{X_i\}_{i=1}^n$ .
- 2) compute

$$\frac{1}{n} \sum_{i=1}^n I(0.3 < X_i < 1.2) \cdot \frac{h(X_i)}{g(X_i)} = \frac{1}{n} \sum_{i=1}^n I(0.3 < X_i < 1.2) \sin^2(X_i) \quad (3)$$

$$\frac{1}{n} \sum_{i=1}^n \frac{h(X_i)}{g(X_i)} = \frac{1}{n} \sum_{i=1}^n \sin^2(X_i) \quad (4)$$

Then (3)/(4)  $\rightarrow P(0.3 < X < 1.2)$  in probability.

### 3

$p(\mathbf{x}) \leq 1$ , let  $h(\mathbf{x}) = I(\mathbf{x} \in [0, 1]^5)$  be the uniform density. It could be simulated by drawing 5 iid  $\text{unif}(0,1)$  coordinates  $x_i, 1 \leq i \leq 5$ .

Note  $1 \cdot h(\mathbf{x}) \geq p(\mathbf{x})$ . To get a single realization  $\mathbf{X}^*$ , do the following:

- 1) generate  $\mathbf{X}^{**} = (X_1^{**}, X_2^{**}, X_3^{**}, X_4^{**}, X_5^{**})$  from  $h$ .
- 2) generate  $U \sim \text{unif}(0, 1)$  which is independent from  $\mathbf{X}^{**}$ .
- 3) if  $Uh(\mathbf{X}^{**}) = U < p(\mathbf{X}^{**})$ , set  $\mathbf{X}^* = \mathbf{X}^{**}$ , otherwise return to 1).