

$$5.27. c) f_{X_{(i)} | X_{(j)}}(u, v) = \frac{f_{X_{(i)}, X_{(j)}}(u, v)}{f_{X_{(j)}}(v)}$$

For  $i < j$ ,

$$= \frac{n!}{(i-1)!(j-i-1)!(n-j)!} f_X(u) f_X(v) [F_X(u)]^{i-1} [F_X(v) - F_X(u)]^{j-i-1} [1-F_X(v)]^{n-j}$$

$$= \frac{n!}{(j-1)!(n-j)!} f_X(v) [F_X(v)]^{j-1} [1-F_X(v)]^{n-j}$$

(3)

$$= \frac{(j-1)!}{(i-1)!(j-i-1)!} \frac{f_X(u)}{f_X(v)} \left[ \frac{F_X(u)}{F_X(v)} \right]^{i-1} \left[ \frac{1-F_X(u)}{1-F_X(v)} \right]^{j-i-1} \quad u < v$$

For  $j < i$ ,  $u > v$  similarly we get

$$f_{X_{(i)} | X_{(j)}}(u, v) = \frac{(n-j)!}{(i-j-1)!(n-i)!} \frac{f_X(u)}{1-F_X(v)} \left[ \frac{F_X(u) - F_X(v)}{1-F_X(v)} \right]^{i-j-1} \left[ \frac{1-F_X(u) - F_X(v)}{1-F_X(v)} \right]^{n-i}$$

b) we have that  $f_{R,V}(r, v) = \frac{n(n-1)}{a^n} r^{n-2} \quad 0 < r < a, \quad \tau/2 < v < a - \tau/2$

$$(2) \quad f_R(r) = \frac{n(n-1)}{a^n} r^{n-2} (a-r) \quad 0 < r < a$$

$$\therefore f_{R|V}(r|v) = \frac{1}{a-r} \quad \tau/2 < v < a - \tau/2$$

5.36 a) We use Theorem 4.4.3 which states that

$$E[X] = E[E[X|Y]]$$

and Theorem 4.4.7 (p 167)

$$\text{Var}[X] = E[\text{Var}(X|Y)] + \text{Var}[E[X|Y]]$$

In this example we have

$$E[Y|N=n] = 2n \quad (\text{mean of } \chi^2_{2n})$$

$$\text{Var}[Y|N=n] = 2(2n) = 4n \quad (\text{Var of } \chi^2_{2n})$$

$$\therefore E[Y] = E[2N] = 2\theta \quad (\text{mean of Poisson } (\theta))$$

$$\text{Var}[E[Y|N]] = \text{Var}[2N] = 4\theta \quad (\text{Variance of Poisson } \theta)$$

$$E[\text{Var}[Y|N]] = E[4N] = 4\theta \quad (\text{mean of Poisson } (\theta))$$

$$\therefore \text{Var}[X] = 8\theta$$

b) We will calculate the moment generating function of

$$Z = \frac{Y - EY}{\sqrt{\text{Var}Y}} = \frac{Y - 2\theta}{\sqrt{8\theta}}$$

The mgf of  $P(\theta)$  is  $e^{\theta(t-1)}$

The mgf of  $\chi^2_{2n}$  is  $(1-2t)^{-n}$

$$M_Z(t) = E[e^{tZ}] = E\left[e^{\frac{t}{\sqrt{8\theta}}Y}\right] \cdot e^{-\left(\frac{2\theta}{\sqrt{8\theta}}\right)t} \quad \text{Set } a = \frac{t}{\sqrt{8\theta}}$$

$$\text{Now } E[e^{atY}] = E E[e^{aY} | N=n] = E\left[(1-2at)^{-N}\right]$$

$$= \sum_{n=0}^{\infty} (1-2at)^{-n} e^{-\theta} \frac{\theta^n}{n!} = e^{-\theta} \exp\left[\frac{\theta}{1-2at}\right]$$

$$\text{Consider } \log M_Z(t) = -\theta + \frac{\theta}{1-2at} - \frac{2\theta}{\sqrt{8\theta}}t \rightarrow \frac{t^2}{2} \text{ Q.E.D.}$$

$$5.42 \quad a) \quad f_X(x) = \beta(1-x)^{\beta-1} \quad 0 < x < 1 \quad \beta > 0$$

$$F_X(x) = 1 - (1-x)^\beta$$

$$\therefore P(X_{(n)} \leq 1-x) = (1-x)^\beta$$

$$\therefore P(1 - X_{(n)} \leq x) = (1-x)^\beta$$

Choose  $x = \frac{\varepsilon}{n^\nu}$ . We know  $(1 + \frac{x}{n})^n \rightarrow e^x$  as  $n \rightarrow \infty$ .

$$\therefore 1 - \frac{\varepsilon^\beta}{n^{\beta\nu}} = 1 - \frac{\varepsilon^\beta}{n} \left( \frac{n}{n^{\beta\nu}} \right)$$

$$\text{Choose } \nu \Rightarrow \frac{n}{n^{\beta\nu}} = 1 \Rightarrow \frac{1}{n^{\beta\nu-1}} = 1 \Rightarrow \beta\nu = 1$$

$\therefore n^{1/\beta}(1 - X_{(n)})$  will have cdf  $e^{-x^\beta}$

$$b) \quad X \sim \text{exp}(1) \Rightarrow f_X(x) = \begin{cases} e^{-x} & , \quad x > 0 \\ 0 & \text{o.w.} \end{cases}$$

$$F(x) = 1 - e^{-x}$$

$$P(X_{(n)} \leq x) = (1 - e^{-x})^n$$

$$\text{Set } x = u + a_n$$

$$\therefore 1 - e^{-x} = 1 - e^{-u} \cdot e^{a_n} = 1 - \frac{e^{-u}}{n} \cdot n e^{a_n}$$

$$\text{Choose } a_n \text{ such that } n e^{a_n} = 1 \Rightarrow a_n = -\log n$$

5.49 a) Trivial

b) Let  $x = g(u) = -\log \frac{1-u}{u}$

Then  $g^{-1}(x) = \frac{1}{1+e^{-x}}$

and  $f_X(x) = \frac{e^{-x}}{(1+e^{-x})^2}$  which is the logistic density  
 $-\infty < x < \infty$

c) If  $X$  is logistic  $(\mu, \beta)$  then

$$f_X(x) = \frac{1}{\beta} f_Y\left(\frac{-x+\mu}{\beta}\right) \text{ where } Y \sim \text{logistic}(0, 1)$$

So to generate  $X$ , first generate  $U \sim \text{Unif}(0, 1)$

and set  $X = \beta \log \frac{U}{1-U} + \mu$ .