## STAT/MATH 415 HW#1

Sep 8, 2017

## **EXERCISES**

5.1.5 The p.d.f of X is  $f(x) = \theta x^{\theta-1}$ , 0 < x < 1,  $0 < \theta < \infty$ . Let  $Y = -2\theta \ln X$ . How is Y distributed? **Answer:** From the support of X, we can get  $0 < Y < \infty$  and X-Y have 1-1 mapping.

$$Y = u(x) = -2\theta \ln X$$
 
$$X = v(y) = e^{-y/2\theta} \quad \text{so} \quad v'(y) = -\frac{1}{2\theta} e^{-y/2\theta}$$
 
$$f_Y(y) = f_X(v(y)) |v'(y)| = \theta (e^{-y/2\theta})^{\theta - 1} \frac{1}{2\theta} e^{-y/2\theta} = \frac{1}{2} e^{-y/2} \quad 0 < y < \infty$$

Thus Y follows the exponential distribution with parameter  $\theta = 2$ .

5.1.15 Let  $Y = X^2$ 

a) Find the p.d.f of Y when the distribution of X is N(0, 1)

**Answer:** The support of Y is  $Y \ge 0$ 

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2} \qquad -\infty < x < \infty$$

$$Y = u(x) = x^2 \qquad X = \begin{cases} v_1(y) = \sqrt{y} & x \ge 0 \\ v_2(y) = -\sqrt{y} & x < 0 \end{cases}$$

$$g_1(y) = f(v_1(y)) |v_1'(y)| = \frac{1}{\sqrt{2\pi}} e^{-y/2} \frac{1}{2\sqrt{y}} \qquad x \ge 0$$

$$g_2(y) = f(v_2(y)) |v_2'(y)| = \frac{1}{\sqrt{2\pi}} e^{-y/2} |-\frac{1}{2\sqrt{y}}| \qquad x < 0$$

$$g(y) = g_1(y) + g_2(y) = \frac{1}{\sqrt{2\pi y}} e^{-y/2} \qquad y \ge 0$$

b) Find the p.d.f of Y when the p.d.f of X is  $f(x) = (3/2)x^2$ , -1 < x < 1

**Answer:** The support of Y is  $0 \le y < 1$ 

$$F_X(x) = \int_{-1}^x f(t)dt = \frac{t^3}{2} \Big|_{-1}^x = \frac{x^3 + 1}{2} - 1 < x < 1$$

$$F_Y(y) = P(Y \le y) = P(x^2 \le y) = P(-\sqrt{y} \le x \le \sqrt{y}) = F_X(\sqrt{y}) - F_X(-\sqrt{y}) = y^{3/2}$$

$$f_Y(y) = F_Y'(y) = \frac{3}{2}\sqrt{y} \qquad 0 \le y < 1$$

5.2.1 Let  $X_1, X_2$  denote two independent random variables, each with a  $\chi^2(2)$  distribution. Find the joint pdf of  $Y_1 = X_1$  and  $Y_2 = X_2 + X_1$ . Note that the support of  $Y_1, Y_2$  is  $0 < y_1 < y_2 < \infty$ . Also, find the marginal pdf of each of  $Y_1$  and  $Y_2$ . Are  $Y_1$  and  $Y_2$  independent?

**Answer:**  $(X_1, X_2) - (Y_1, Y_2)$  is 1-1 mapping and  $\chi^2(2) = \gamma(1, 2)$ 

$$f(x;1,2) = \frac{x^{1-1}e^{-x/2}}{2^{1}\Gamma(1)} = \frac{e^{-x/2}}{2} \qquad \Rightarrow \qquad f(x_{1},x_{2}) = f(x_{1})f(x_{2}) = \frac{1}{4}e^{-(x_{1}+x_{2})/2}$$

$$x_{1} = v_{1}(y_{1},y_{2}) = y_{1} \qquad x_{2} = v_{2}(y_{1},y_{2}) = y_{2} - y_{1} \qquad \Rightarrow \qquad J = \begin{vmatrix} \frac{\partial x_{1}}{\partial y_{1}} & \frac{\partial x_{1}}{\partial y_{2}} \\ \frac{\partial x_{2}}{\partial y_{1}} & \frac{\partial x_{2}}{\partial y_{2}} \end{vmatrix} = \begin{vmatrix} 1 & 0 \\ -1 & 1 \end{vmatrix} = 1$$

$$g(y_{1},y_{2}) = f(v_{1},v_{2})|J| = \frac{1}{4}e^{-y_{2}/2} \qquad y_{2} > 0 \qquad \Rightarrow \qquad g_{1}(y_{1}) = \int_{y_{1}}^{\infty} g(y_{1},y_{2})dy_{2} = \frac{1}{2}e^{-y_{1}/2} \qquad y_{1} > 0$$

$$\Rightarrow \qquad g_{2}(y_{2}) = \int_{0}^{y_{2}} g(y_{1},y_{2})dy_{1} = \frac{y_{2}}{4}e^{-y_{2}/2} \qquad y_{2} > 0$$

$$g(y_{1},y_{2}) \neq g_{1}(y_{1})g_{2}(y_{2}) \qquad \Rightarrow \qquad Y_{1} \text{ and } Y_{2} \text{ are not independent}$$

5.2.5 Let the distribution of W be F(8, 4). Find the following

a)  $F_{0.01}(8,4)$ 

**Answer:** That is  $0.01 = P(W \ge F_{0.01}(8,4))$ . According to the appendix table, we can get W = 14.80

b) 
$$F_{0.99}(8,4)$$

**Answer:** 
$$F_{1-0.01}(8,4) = \frac{1}{F_{0.01}(4,8)}$$
  $\Rightarrow$   $F_{0.99}(8,4) = \frac{1}{7.01} \approx 0.1427$ 

c)  $P(0.198 \le W \le 8.98)$ 

**Answer:** From the table, we can get  $F_{0.025}(8,4) = 8.98, F_{0.025}(4,8) = \frac{1}{0.108}$ 

$$0.198 = \frac{1}{F_{0.025}(4,8)} = F_{1-0.025}(8,4) \qquad \Rightarrow \qquad P(W \le F_{1-0.025}(8,4)) = 0.025$$
$$8.98 = F_{0.025}(8,4) \qquad \Rightarrow \qquad P(W \le F_{0.025}(8,4)) = 0.975$$

Thus,  $P(0.198 \le W \le 8.98) = 0.975 - 0.025 = 0.950$ 

5.2.6 Let  $X_1$  and  $X_2$  have independent gamma distributions with parameters  $\alpha, \theta$  and  $\beta, \theta$  respectively. Let  $W = X_1/(X_1 + X_2)$ . Use a method similar to that given in the derivation of F distribution to show that the pdf of W is

$$g(w) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} w^{\alpha - 1} (1 - w)^{\beta - 1}, \quad 0 < w < 1$$

We say that W has a beta distribution with parameters  $\alpha$  and  $\beta$ 

**Answer:**  $X_1 \sim \Gamma(\alpha, \theta)$  and  $X_2 \sim \Gamma(\beta, \theta)$ . Let  $U = X_1 + X_2$ , then

$$x_1 = v_1(u, w) = uw \qquad x_2 = v_2(u, w) = u(1 - w) \qquad \Rightarrow \qquad J = \begin{vmatrix} \frac{\partial v_1}{\partial u} & \frac{\partial v_1}{\partial w} \\ \frac{\partial v_2}{\partial u} & \frac{\partial v_2}{\partial w} \end{vmatrix} = \begin{vmatrix} w & u \\ 1 - w & -u \end{vmatrix} = -u$$

The support of W is (0, 1), of U is  $(0, \infty)$ 

$$\begin{split} f_1(x_1;\alpha,\theta) &= \frac{x_1^{\alpha-1}e^{-x_1/\theta}}{\theta^{\alpha}\Gamma(\alpha)} \quad x_1 > 0 \quad \text{ and } \quad f_2(x_2;\beta,\theta) = \frac{x_2^{\beta-1}e^{-x_2/\theta}}{\theta^{\beta}\Gamma(\beta)} \quad x_2 > 0 \\ f(x_1,x_2) &= \frac{x_1^{\alpha-1}x_2^{\beta-1}e^{-(x_1+x_2)/\theta}}{\theta^{\alpha+\beta}\Gamma(\alpha)\Gamma(\beta)} \quad x_1,x_2 > 0 \\ g(u,w) &= f(v_1(u,w),v_2(u,w))\big|J\big| = \frac{(uw)^{\alpha-1}(u(1-w))^{\beta-1}e^{-u/\theta}}{\theta^{\alpha+\beta}\Gamma(\alpha)\Gamma(\beta)} u \\ &= \frac{u^{\alpha+\beta-1}w^{\alpha-1}(1-w)^{\beta-1}e^{-u/\theta}}{\theta^{\alpha+\beta}\Gamma(\alpha)\Gamma(\beta)} \quad u > 0, \ 0 < w < 1 \\ g(w) &= \int_0^\infty g(u,w)du \\ &= \frac{w^{\alpha-1}(1-w)^{\beta-1}}{\theta^{\alpha+\beta}\Gamma(\alpha)\Gamma(\beta)} \int_0^\infty u^{\alpha+\beta-1}e^{-u/\theta}du \\ \text{Note that } \frac{u^{\alpha+\beta-1}e^{-u/\theta}}{\Gamma(\alpha+\beta)\theta^{\alpha+\beta}} \text{ is the form of gamma distribution} \\ &= \frac{w^{\alpha-1}(1-w)^{\beta-1}}{\theta^{\alpha+\beta}\Gamma(\alpha)\Gamma(\beta)}\Gamma(\alpha+\beta)\theta^{\alpha+\beta} = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)}w^{\alpha-1}(1-w)^{\beta-1}, \quad 0 < w < 1 \end{split}$$

Thus W has a beta distribution with parameters  $\alpha$  and  $\beta$ 

5.3.4 Let  $X_1$  and  $X_2$  be a random sample of size n = 2 from the exponential distribution with pdf  $f(x) = 2e^{-2x}$ ,  $0 < x < \infty$ . Find

a) 
$$P(0.5 < X_1 < 1.0, 0.7 < X_2 < 1.2)$$

**Answer:**  $X_1 \sim Exp(2)$ ,  $X_2 \sim Exp(2)$  and we can assume that  $X_1, X_2$  are independent.

$$\begin{split} F(x) &= 1 - e^{-2x}, \ x \geq 0 \\ P(0.5 < X_1 < 1.0) &= F(1.0) - F(0.5) = e^{-1} - e^{-2} = 0.2325 \\ P(0.7 < X_2 < 1.2) &= F(1.2) - F(0.7) = e^{-1.4} - e^{-2.4} = 0.1559 \\ P(0.5 < X_1 < 1.0, \ 0.7 < X_2 < 1.2) &= P(0.5 < X_1 < 1.0) P(0.7 < X_2 < 1.2) = 0.0362 \end{split}$$

b) 
$$E[X_1(X_2 - 0.5)^2]$$
  
Answer:  $E(X_1) = E(X_2) = \frac{1}{2}$ .  $V(X_1) = V(X_2) = \frac{1}{4}$   
 $E[X_1(X_2 - 0.5)^2] = E(X_1)E[(X_2 - 0.5)^2] = E(X_1)V(X_2)$   
 $= 0.5(0.25) = 0.125$ 

5.3.11 Let  $X_1, X_2, X_3$  be three independent random variables with binomial distributions b(4, 1/2), b(6, 1/3), b(12, 1/6) respectively. Find

a) 
$$P(X_1 = 2, X_2 = 2, X_3 = 5)$$

**Answer:** Since  $X_1, X_2, X_3$  are independent, then

$$P(X_1 = 2, X_2 = 2, X_3 = 5) = P(X_1 = 2)P(X_2 = 2)P(X_3 = 5)$$

$$= {4 \choose 2} (\frac{1}{2})^2 (\frac{1}{2})^2 \times {6 \choose 2} (\frac{1}{3})^2 (\frac{2}{3})^4 \times {12 \choose 5} (\frac{1}{6})^5 (\frac{5}{6})^7$$

$$= \frac{6}{2^4} \times \frac{15 \times 16}{3^6} \times \frac{792 \times 5^7}{6^{12}} = 0.0035$$

b)  $E(X_1X_2X_3)$ 

**Answer:** Due to their independence, we can get  $E(X_1X_2X_3) = E(X_1)E(X_2)E(X_3) = 2 \times 2 \times 2 = 8$ 

c) The mean and the variance of  $Y = X_1 + X_2 + X_3$ 

**Answer:** Based on  $Var(X) = E(X^2) - E(X)^2$ 

$$E(X_1^2) = 1 + 2^2 = 5 E(X_2^2) = \frac{4}{3} + 2^2 = \frac{16}{3} E(X_3^2) = \frac{5}{3} + 2^2 = \frac{17}{3}$$

$$E(Y^2) = E(X_1^2 + X_2^2 + X_3^2 + 2X_1X_2 + 2X_1X_3 + 2X_2X_3)$$

$$= E(X_1^2) + E(X_2^2) + E(X_3^2) + 2E(X_1X_2) + 2E(X_1X_3) + 2E(X_2X_3)$$

$$= 5 + \frac{16}{3} + \frac{17}{3} + 8 + 8 + 8 = 40$$

$$E(Y) = E(X_1) + E(X_2) + E(X_3) = 2 + 2 + 2 = 6$$

$$Var(Y) = E(Y^2) - E(Y)^2 = 40 - 6^2 = 4$$