

HW 7 solution

- 6.10** The total time from arrival to completion of service at a fast-food outlet, Y_1 , and the time spent waiting in line before arriving at the service window, Y_2 , were given in Exercise 5.15 with joint density function

$$f(y_1, y_2) = \begin{cases} e^{-y_1}, & 0 \leq y_2 \leq y_1 < \infty, \\ 0, & \text{elsewhere.} \end{cases}$$

Another random variable of interest is $U = Y_1 - Y_2$, the time spent at the service window. Find

- a the probability density function for U .
- b $E(U)$ and $V(U)$. Compare your answers with the results of Exercise 5.108.

$$\text{a. } F(u) = P(Y_1 - Y_2 \leq u) = \int_0^{\infty} \int_{y_2}^{u+y_2} f(y_1, y_2) dy_1 dy_2 = 1 - e^{-u}$$

$$\therefore f'(u) = f(u) = e^{-u} \quad (u \geq 0)$$

$$\text{b. } E(V) = \int_0^{\infty} u \cdot e^{-u} du = 1$$

$$V(V) = \int_0^{\infty} u^2 \cdot e^{-u} du - E(V)^2 = 1 \quad \rightarrow V \sim \exp(1) \quad \text{then } E(V) = 1, V(V) = 1$$

- 6.12** Suppose that Y has a gamma distribution with parameters α and β and that $c > 0$ is a constant.

- a Derive the density function of $U = cY$.
- b Identify the density of U as one of the types we studied in Chapter 4. Be sure to identify any parameter values.
- c The parameters α and β of a gamma-distributed random variable are, respectively, "shape" and "scale" parameters. How do the scale and shape parameters for U compare to those for Y ?

(a), (b)

$$f_Y(y) = \frac{1}{\Gamma(\alpha) \cdot \beta^\alpha} \cdot y^{\alpha-1} \cdot e^{-y/\beta}$$

$$F_U(u) = P(cY \leq u) = P(Y \leq u/c), \quad f_U(u) = F'(u/c) = \frac{1}{c} f_Y(u/c) = \frac{1}{c} \cdot \frac{1}{\Gamma(\alpha) \cdot \beta^\alpha} \cdot \left(\frac{u}{c}\right)^{\alpha-1} \cdot \exp\left(-\frac{u}{c\beta}\right)$$

$$\therefore f_U(u) = \frac{1}{\Gamma(\alpha) (c\beta)^\alpha} \cdot u^{\alpha-1} \cdot \exp\left(-\frac{u}{c\beta}\right) \quad (u \geq 0) \sim \text{gamma}(\alpha, c\beta)$$

(c) The shape parameter is the same and the scale parameter is c times beta

- 6.17** A member of the power family of distributions has a distribution function given by

(5pts)

$$F(y) = \begin{cases} 0, & y < 0, \\ \left(\frac{y}{\theta}\right)^\alpha, & 0 \leq y \leq \theta, \\ 1, & y > \theta, \end{cases}$$

where $\alpha, \theta > 0$.

- a Find the density function.
- b For fixed values of α and θ , find a transformation $G(U)$ so that $G(U)$ has a distribution function of F when U possesses a uniform $(0, 1)$ distribution.
- c Given that a random sample of size 5 from a uniform distribution on the interval $(0, 1)$ yielded the values .2700, .6901, .1413, .1523, and .3609, use the transformation derived in part (b) to give values associated with a random variable with a power family distribution with $\alpha = 2, \theta = 4$.

a. $f(y) = \frac{\alpha}{\theta} \left(\frac{y}{\theta}\right)^{\alpha-1}$ ($0 \leq y \leq \theta$) \rightarrow easy to check

b. $Y = G(U)$, $U \sim \text{Unif}(0, 1) \rightarrow f_U(u) = 1$

$$f_Y(y) = f_U(G^{-1}(y)) \cdot \left| \frac{d}{dy} G^{-1}(y) \right| = \frac{\alpha}{\theta^\alpha} \cdot y^{\alpha-1} \quad (\text{result of (a)}), \quad 0 \leq y \leq \theta, \quad 0 \leq u \leq 1$$

$$= \frac{1}{\theta} G^{-1}(y) = \frac{\alpha}{\theta^\alpha} \cdot y^{\alpha-1}$$

$$\therefore G^{-1}(y) = \frac{y^\alpha}{\theta^\alpha} = u \quad \Leftrightarrow \quad Y = G(u) = \theta \cdot u^{\frac{1}{\alpha}}$$

c. $Y = 4\sqrt{u}$, calculate by yourself.

6.28 Let Y have a uniform (0, 1) distribution. Show that $U = -2 \ln(Y)$ has an exponential distribution with mean 2.

$$F_U(u) = P(-2 \ln(Y) \leq u) = P(Y \geq e^{-\frac{u}{2}}) = 1 - F_Y(e^{-\frac{u}{2}})$$

$$\therefore f_U(u) = (F_U(u))' = \frac{1}{2} e^{-\frac{u}{2}} \quad (u \geq 0). \quad \text{pdf of exponential dist with beta=2}$$

$$\therefore E(U) = 2$$

6.31 The joint distribution for the length of life of two different types of components operating in a system was given in Exercise 5.18 by

$$f(y_1, y_2) = \begin{cases} (1/8)y_1 e^{-(y_1+y_2)/2}, & y_1 > 0, y_2 > 0, \\ 0, & \text{elsewhere.} \end{cases}$$

The relative efficiency of the two types of components is measured by $U = Y_2/Y_1$. Find the probability density function for U .

Let $U = Y_2/Y_1$, $V = Y_1 \Rightarrow Y_1 = V$

$$f_{U,V}(u,v) = f_{Y_1, Y_2}(y_1, y_2) \cdot |J|, \quad |J| = \begin{vmatrix} 1 & 0 \\ 0 & v \end{vmatrix} = v$$

$$= \frac{1}{8} \cdot v^2 \cdot e^{-v(1+v)/2}$$

$$f(v) = \int_{-\infty}^{\infty} f_{U,V}(v,u) du = \int_0^{\infty} \frac{1}{8} v^2 \cdot e^{-v(1+u)/2} du = \dots = \frac{2}{(1+u)^3}$$

calculate yourself

6.43 Refer to Exercise 6.41. Let Y_1, Y_2, \dots, Y_n be independent, normal random variables, each with mean μ and variance σ^2 .

a Find the density function of $\bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i$.

b If $\sigma^2 = 16$ and $n = 25$, what is the probability that the sample mean, \bar{Y} , takes on a value that is within one unit of the population mean, μ ? That is, find $P(|\bar{Y} - \mu| \leq 1)$.

c If $\sigma^2 = 16$, find $P(|\bar{Y} - \mu| \leq 1)$ if $n = 36$, $n = 64$, and $n = 81$. Interpret the results of your calculations.

$$a. \bar{Y} \sim N\left(0, \frac{\sigma^2}{n}\right)$$

\Rightarrow we learned exactly the same example using mgf in class.

$$b. P(|\bar{Y}-0| \leq 1) \Leftrightarrow P\left(|Z| \leq \frac{5}{4}\right) \approx 0.7888$$

$$c. P(|\bar{Y}-0| \leq 1) \Leftrightarrow P\left(|Z| \leq \frac{\sqrt{n}}{6}\right) = P\left(|Z| \leq \frac{\sqrt{n}}{4}\right)$$

$$n=36 \Rightarrow P(|Z| \leq 1.5) \approx 0.8664$$

$$n=64 \Rightarrow P(|Z| \leq 2) \approx 0.9544$$

$$n=81 \Rightarrow P(|Z| \leq 2.25) \approx 0.9756$$

$\rightarrow n$ gets large, probability also large

- 6.46 Suppose that Y has a gamma distribution with $\alpha = n/2$ for some positive integer n and β equal to some specified value. Use the method of moment-generating functions to show that $W = 2Y/\beta$ has a χ^2 distribution with n degrees of freedom.

$$M_Y(t) = (1-\beta t)^{-\frac{n}{2}}$$

$$\begin{aligned} \text{mgf of } W &= M_W(t) = E(e^{tW}) = E\left(e^{\frac{2tY}{\beta}}\right) \\ &= (1 - \beta \frac{2t}{\beta})^{-\frac{n}{2}} = (1 - 2t)^{-\frac{n}{2}} \Rightarrow \text{mgf of } \chi^2(n) \end{aligned}$$

- 6.60 Suppose that $W = Y_1 + Y_2$ where Y_1 and Y_2 are independent. If W has a χ^2 distribution with v degrees of freedom and W_1 has a χ^2 distribution with $v_1 < v$ degrees of freedom, show that Y_2 has a χ^2 distribution with $v - v_1$ degrees of freedom.

$$M_W(t) = M_{Y_1}(t) \cdot M_{Y_2}(t) \quad , \quad M_{Y_2}(t) = \frac{M_w(t)}{M_{Y_1}(t)} = \frac{(1-2t)^{-\frac{v}{2}}}{(1-2t)^{-\frac{v_1}{2}}} = (1-2t)^{-\frac{(v-v_1)}{2}}$$

$$\therefore Y_2 \sim \chi^2(v-v_1)$$

- *6.62 Let Y_1 and Y_2 be independent normal random variables, each with mean 0 and variance σ^2 . Define $U_1 = Y_1 + Y_2$ and $U_2 = Y_1 - Y_2$. Show that U_1 and U_2 are independent normal random variables, each with mean 0 and variance $2\sigma^2$. [Hint: If (U_1, U_2) has a joint moment-generating function $m(t_1, t_2)$, then U_1 and U_2 are independent if and only if $m(t_1, t_2) = m_{U_1}(t_1)m_{U_2}(t_2)$.]

$$M_{U_1, U_2}(t_1, t_2) = E(e^{t_1 U_1 + t_2 U_2}) = E(e^{t_1(Y_1+Y_2) + t_2(Y_1-Y_2)})$$

$$= M_{Y_1}(t_1+t_2) \cdot M_{Y_2}(t_1-t_2), \text{ mgf of } M_{Y_1}(t) = \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right) \xrightarrow{\text{mgf of }} N(0, 2\sigma^2)$$

$$\Rightarrow \exp\left(\frac{\sigma^2}{2}(t_1+t_2)^2\right) \cdot \exp\left(\frac{\sigma^2}{2}(t_1-t_2)^2\right) = \exp\left(\frac{\sigma^2 t_1^2}{2}\right) \cdot \exp\left(\frac{\sigma^2 t_2^2}{2}\right) = M_{U_1}(t_1) \cdot M_{U_2}(t_2)$$

6.65 Let Z_1 and Z_2 be independent standard normal random variables and $U_1 = Z_1$ and $U_2 = Z_1 + Z_2$. (10pts)

- a Derive the joint density of U_1 and U_2 .
- b Use Theorem 5.12 to give $E(U_1)$, $E(U_2)$, $V(U_1)$, $V(U_2)$, and $\text{Cov}(U_1, U_2)$.
- c Are U_1 and U_2 independent? Why?
- d Refer to Section 5.10. Show that U_1 and U_2 have a bivariate normal distribution. Identify all the parameters of the appropriate bivariate normal distribution.

$$a. f_{U_1, U_2}(u_1, u_2) = f_{Z_1, Z_2}(z_1, z_2) \cdot |\mathcal{J}|, |\mathcal{J}| = 1$$

$$\Rightarrow \frac{1}{2\pi} \cdot \exp^{-\frac{(u_1^2 + (u_2 - u_1)^2)}{2}}$$

$$b. E(U_1) = 0, E(U_2) = E(Z_1 + Z_2) = 0, V(U_1) = 1, V(U_2) = V(Z_1 + Z_2) = 2$$

$$\text{Cov}(U_1, U_2) = \text{Cov}(Z_1, Z_1 + Z_2) = \text{Var}(Z_1) + \text{Cov}(Z_1, Z_2) = 1$$

c. by (b), $\text{Cov}(U_1, U_2) \neq 0$ Hence U_1, U_2 is not independent.

d. see pdf of bivariate normal distribution, $\begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \Sigma = \begin{pmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & 2 \end{pmatrix}$

$$\Rightarrow \sigma_1^2 = 1, \sigma_2^2 = 2, \rho = \frac{1}{\sqrt{2}}$$
(5pts)

***6.68** Let Y_1 and Y_2 have joint density function

$$f_{Y_1, Y_2}(y_1, y_2) = \begin{cases} 8y_1y_2, & 0 \leq y_1 < y_2 \leq 1, \\ 0, & \text{otherwise,} \end{cases}$$

and $U_1 = Y_1/Y_2$ and $U_2 = Y_2$.

- a Derive the joint density function for (U_1, U_2) .
- b Show that U_1 and U_2 are independent.

$$a. f_{U_1, U_2}(u_1, u_2) = f_{Y_1, Y_2}(y_1, y_2) |\mathcal{J}| = 8(u_1 u_2) \cdot u_2 \cdot |u_2| = 8u_1 u_2^3$$

$$(0 \leq u_1, u_2 \leq 1)$$

b. We need to show $f_{U_1}(u_1) \cdot f_{U_2}(u_2) = f_{U_1, U_2}(u_1, u_2)$

$$f_{U_1}(u_1) = \int_0^1 8u_1 u_2^3 du_2 = 2u_1,$$

$$f_{U_2}(u_2) = \int_0^1 8u_1 u_2^3 du_1 = 4u_2^3, \quad \therefore f_{U_1}(u_1) \cdot f_{U_2}(u_2) = f_{U_1, U_2}(u_1, u_2)$$

Hence U_1 and U_2 are independent.

***6.71** Suppose that Y_1 and Y_2 are independent exponentially distributed random variables, both with mean β , and define $U_1 = Y_1 + Y_2$ and $U_2 = Y_1/Y_2$.

(5pts)

- a Show that the joint density of (U_1, U_2) is

$$f_{U_1, U_2}(u_1, u_2) = \begin{cases} \frac{1}{\beta^2} u_1 e^{-u_1/\beta} \frac{1}{(1+u_2)^2}, & 0 < u_1, 0 < u_2, \\ 0, & \text{otherwise.} \end{cases}$$

- b Are U_1 and U_2 independent? Why?

$$a. f_{U_1, U_2}(u_1, u_2) = f_{Y_1, Y_2}(y_1, y_2) |J| = f_{Y_1}(y_1) \cdot f_{Y_2}(y_2) \cdot |J|$$

$$|J| = \left| \frac{-\alpha_1}{(1+\alpha_2)^2} \right|, \quad f_{U_1, U_2}(u_1, u_2) = \frac{1}{\beta^2} \cdot \exp\left(-\frac{u_1}{\beta}\right) \cdot \frac{\alpha_1}{(1+\alpha_2)^2}, \quad u_1, u_2 > 0$$

$$b. f_{U_2}(u_2) = \int_0^\infty \frac{1}{\beta^2} \cdot \exp\left(-\frac{u_1}{\beta}\right) \cdot \frac{\alpha_1}{(1+\alpha_2)^2} du_1 = \frac{1}{\beta^2} e^{-\frac{u_1}{\beta}} \cdot u_1,$$

$$f_{U_1}(u_1) = \int_0^\infty \frac{1}{\beta^2} \cdot \exp\left(-\frac{u_1}{\beta}\right) \cdot \frac{\alpha_1}{(1+\alpha_2)^2} du_2 = \frac{1}{(1+\alpha_2)^2}$$

$\therefore f_{U_1}(u_1) f_{U_2}(u_2) = f_{U_1, U_2}(u_1, u_2)$, U_1 and U_2 are independent.

- 6.79 Refer to Exercise 6.77. If Y_1, Y_2, \dots, Y_n are independent, uniformly distributed random variables on the interval $[0, \theta]$, show that $U = Y_{(1)}/Y_{(n)}$ and $Y_{(n)}$ are independent.

$$\text{Let } V = Y_{(n)}, \quad f_{U, V}(u, v) = f_{Y_{(1)}, Y_{(n)}}(uv, v) |J|$$

$$f_{Y_{(1)}, Y_{(n)}}(y_1, y_n) = \frac{n!}{(n-1)! (n-2)! \cdots 1!} \cdot [F(y_n)]^{n-1} f(y_n) \cdot [F(y_n) - F(y_1)]^{n-2} f(y_1) \cdot [1 - F(y_1)]^{n-1}$$

(see lecture note)

$$\therefore f_{Y_{(1)}, Y_{(n)}}(uv, v) |J| = n(n-1) \left(\frac{1}{\theta}\right)^n (v-uv)^{n-2} |v| = n(n-1) \left(\frac{1}{\theta}\right)^n (1-u)^{n-2} v^{n-2}$$

$(0 \leq u \leq 1, 0 \leq v \leq \theta)$

$$f_U(u) \cdot f_V(v) = f_{U, V}(u, v) \quad (U \text{ and } V \text{ are independent})$$

- 6.80 Let Y_1, Y_2, \dots, Y_n be independent random variables, each with a beta distribution, with $\alpha = \beta = 2$. Find (5pts)

a the probability distribution function of $Y_{(n)} = \max(Y_1, Y_2, \dots, Y_n)$.

b the density function of $Y_{(n)}$.

c $E(Y_{(n)})$ when $n = 2$.

$$f(y) = 6 \cdot y \cdot (1-y), \quad F(y) = 3y^2 - 2y^3 \quad (0 \leq y \leq 1)$$

$$a. P(Y_{(n)} \leq y) = P(Y_1, Y_2, \dots, Y_n \leq y) = [P(Y_i \leq y)]^n = [F(y)]^n$$

$$= [3y^2 - 2y^3]^n$$

$$b. f_{(n)}(y) = n F(y)^{n-1} \cdot f(y) = 6ny[3y^2 - 2y^3]^{n-1} \cdot (1-y)$$

$$c. f_{(2)}(y) = 12y (3y^2 - 2y^3), \quad \int_0^1 y \cdot 12y (3y^2 - 2y^3) (1-y) dy$$

$$= \frac{36}{5} - 10 + \frac{24}{7} \approx 0.6286$$

- 6.85** Let Y_1 and Y_2 be independent and uniformly distributed over the interval $(0, 1)$. Find $P(2Y_{(1)} < Y_{(2)})$.

$$f_{Y_{(1)}, Y_{(2)}}(y_1, y_2) = 2, \quad (0 \leq y_1 \leq y_2 \leq 1)$$

$$\therefore P(2Y_{(1)} < Y_{(2)}) = \int_0^1 \int_{2y_1}^1 2 dy_2 dy_1 = \frac{1}{2}$$

- 6.87** The opening prices per share Y_1 and Y_2 of two similar stocks are independent random variables, each with a density function given by

$$f(y) = \begin{cases} (1/2)e^{-(1/2)(y-4)}, & y \geq 4, \\ 0, & \text{elsewhere.} \end{cases}$$

On a given morning, an investor is going to buy shares of whichever stock is less expensive. Find the

- a probability density function for the price per share that the investor will pay.
- b expected cost per share that the investor will pay.

$$a. f_{C_1}(y) = n [1 - F(y)]^{n-1} f(y) = e^{-(y-4)} \quad (y \geq 4)$$

$$b. \int_4^\infty y \cdot e^{-(y-4)} dy = 5$$

- *6.103** Let Y_1 and Y_2 be independent, standard normal random variables. Find the probability density function of $U = Y_1/Y_2$.

$$\text{Let } V = Y_2, \quad f_{U,V}(u,v) = f_{Y_1,Y_2}(uv, v) |J| = \frac{1}{2\pi} e^{-\frac{(v^2+u^2)}{2}} |V|$$

$$\therefore f_U(u) = \int_0^\infty \frac{1}{\pi} \exp\left(-\frac{v^2+u^2}{2}\right) dv = \frac{1}{\pi(1+u^2)} \quad (-\infty < u < \infty)$$