Stat 345/445: Theoretical Statistics I: Homework 6 Solutions

Textbook Exercises

4.7 (345: 2 pts & 445: 1 pt.) A woman leaves for work between 8 AM and 8:30 AM and takes between 40 and 50 minutes to get there. Let the random variable X denote her time of departure, and the random variable Y the travel time. Assuming that these variables are independent and uniformly distributed, find the probability that the woman arrives at work before 9 AM.

We will measure time in minutes past 8 AM. So $X \sim \text{uniform}(0,30)$, $Y \sim \text{uniform}(40,50)$ and the joint pdf is 1/300 on the rectangle $(0,30) \times (40,50)$.

$$P(\text{arrive before 9 AM}) = P(X + Y < 60) = \int_{40}^{50} \int_{0}^{60-y} \frac{1}{300} dx dy = \frac{60y - \frac{y^2}{2}}{300} \Big|_{40}^{50}$$
$$= \frac{60(50 - 40) - \frac{(50^2 - 40^2)}{2}}{300} = \frac{600 - 450}{300} = \frac{1}{2}$$

4.10 (345 & 445: 2 pts.) The random pair (X,Y) has the distribution

(a) Show that X and Y are dependent.

The marginal distribution of X is $P(X=1)=P(X=3)=\frac{1}{4}$ and $P(X=2)=\frac{1}{2}$. The marginal distribution of Y is $P(Y=2)=P(Y=3)=P(Y=4)=\frac{1}{3}$. But

$$P(X = 2, Y = 3) = 0 \neq (\frac{1}{2})(\frac{1}{3}) = P(X = 2)P(Y = 3).$$

Therefore the random variables are not independent.

(b) Give a probability table for random variables U and V that have the same marginals as X and Y but are independent.

The distribution that satisfies P(U=x,V=y)=P(U=x)P(V=y) where $U\sim X$ and $V\sim Y$ is

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4.19

(a) (345 & 445: 1 pt.) Let X_1 and X_2 be independent n(0,1) random variables. Find the pdf of $(X_1 - X_2)^2/2$.

Since $(X_1 - X_2)/\sqrt{2} \sim n(0, 1), (X_1 - X_2)^2/2 \sim \mathcal{X}_1^2$ (see Example 2.1.9).

$$Y = \frac{(X_1 - X_2)^2}{2} \sim \mathcal{X}_1^2, \quad f_Y(y) = \frac{1}{\sqrt{2\pi y}} e^{-\frac{y}{2}}, \quad 0 < y < \infty$$

(b) (445: 2 pts.) If X_i , i = 1, 2, are independent gamma(α_i , 1) random variables, find the marginal distributions of $X_1/(X_1 + X_2)$ and $X_2/(X_1 + X_2)$.

Make the transformation $y_1 = \frac{x_1}{x_1 + x_2}$, $y_2 = x_1 + x_2$ then $x_1 = y_1 y_2$, $x_2 = y_2 (1 - y_1)$ and $|J| = y_2$. Then

$$f(y_1, y_2) = \left[\frac{\Gamma(\alpha_1 + \alpha_2)}{\Gamma(\alpha_1)\Gamma(\alpha_2)} y_1^{\alpha_1 - 1} (1 - y_1)^{\alpha_2 - 1} \right] \left[\frac{1}{\Gamma(\alpha_1 + \alpha_2)} y_2^{\alpha_1 + \alpha_2 - 1} e^{-y_2} \right],$$

thus $Y_1 \sim \text{beta}(\alpha_1, \alpha_2), Y_2 \sim \text{gamma}(\alpha_1 + \alpha_2, 1)$ and are independent.

4.22 (345: 1 pt.) Let (X, Y) be a bivariate random vector with joint pdf f(x, y). Let U = aX + b and V = cY + d, where a, b, c, and d are fixed constants with a > 0 and c > 0. Show that the joint pdf of (U, V) is

$$f_{U,V}(u,v) = \frac{1}{ac} f(\frac{u-b}{a}, \frac{v-d}{c}).$$

 $u = ax + b \implies x = \frac{u - b}{a}, \quad v = cy + d \implies y = \frac{v - d}{c}$

$$J = \begin{vmatrix} \frac{dx}{du} & \frac{dx}{dv} \\ \frac{dy}{du} & \frac{dy}{dv} \end{vmatrix} = \begin{vmatrix} \frac{1}{a} & 0 \\ 0 & \frac{1}{c} \end{vmatrix} = \frac{1}{ac}$$

Thus $J = \frac{1}{ac}$.

$$f_{uv}(uv) = f_{xy}(h_x(u)h_y(v))|J| = \frac{1}{ac}f(\frac{u-b}{a} \frac{v-d}{c})$$

- **4.23** For X and Y as in Example 4.3.3, find the distribution of XY by making the transformations given in (a) and integrating out V.
 - (a) (345 & 445: 2 pts.) U = XY, V = Y

Let y = v, x = u/y = u/v then

$$J = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial x}{\partial v} \\ \frac{\partial y}{\partial u} & \frac{\partial y}{\partial v} \end{vmatrix} = \begin{vmatrix} \frac{1}{v} & -\frac{u}{v^2} \\ 0 & 1 \end{vmatrix} = \frac{1}{v}.$$

$$f_{U,V}(u,v) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \left(\frac{u}{v}\right)^{\alpha-1} \left(1 - \frac{u}{v}\right)^{\beta-1} \frac{\Gamma(\alpha+\beta+\gamma)}{\Gamma(\alpha+\beta)\Gamma(\gamma)} v^{\alpha+\beta-1} (1-v)^{\gamma-1} \left(\frac{1}{v}\right)$$

$$= \frac{\Gamma(\alpha+\beta+\gamma)}{\Gamma(\alpha)\Gamma(\beta)\Gamma\gamma} u^{\alpha-1} v^{1-\alpha} \left(1 - \frac{u}{v}\right)^{\beta-1} v^{\alpha+\beta-1} (1-v)^{\gamma-1} v^{-1}$$

$$= \frac{\Gamma(\alpha+\beta+\gamma)}{\Gamma(\alpha)\Gamma(\beta)\Gamma\gamma} u^{\alpha-1} v^{1-\alpha+\alpha+\beta-1-1} \left(1 - \frac{u}{v}\right)^{\beta-1} (1-v)^{\gamma-1}$$

$$= \frac{\Gamma(\alpha+\beta+\gamma)}{\Gamma(\alpha)\Gamma(\beta)\Gamma\gamma} u^{\alpha-1} v^{\beta-1} \left(1 - \frac{u}{v}\right)^{\beta-1} (1-v)^{\gamma-1}$$

$$= \frac{\Gamma(\alpha+\beta+\gamma)}{\Gamma(\alpha)\Gamma(\beta)\Gamma\gamma} u^{\alpha-1} (v-u)^{\beta-1} (1-v)^{\gamma-1}$$

$$0 < u < v < 1.$$

Then,

$$f_U(u) = \int_u^1 \frac{\Gamma(\alpha + \beta + \gamma)}{\Gamma(\alpha)\Gamma(\beta)\Gamma(\gamma)} u^{\alpha - 1} (v - u)^{\beta - 1} (1 - v)^{\gamma - 1} dv$$

Let $z = \frac{v-u}{1-u} \implies v = (1-u)z + u$, dv = (1-u)dz. For v = 1, z = 1 and for v = u, z = 0.

$$f_{U}(u) = \frac{\Gamma(\alpha + \beta + \gamma)}{\Gamma(\alpha)\Gamma(\beta)\Gamma(\gamma)}u^{\alpha - 1} \int_{0}^{1} \left((1 - u)z + u - u \right)^{\beta - 1} \left(1 - (1 - u)z - u \right)^{\gamma - 1} (1 - u)dz$$

$$= \frac{\Gamma(\alpha + \beta + \gamma)}{\Gamma(\alpha)\Gamma(\beta)\Gamma(\gamma)}u^{\alpha - 1} \int_{0}^{1} (1 - u)^{\beta - 1}z^{\beta - 1} (1 - u)^{\gamma - 1} (1 - z)^{\gamma - 1} (1 - u)dz$$

$$= \frac{\Gamma(\alpha + \beta + \gamma)}{\Gamma(\alpha)\Gamma(\beta)\Gamma(\gamma)}u^{\alpha - 1} (1 - u)^{\beta - 1 + \gamma - 1 + 1} \frac{\Gamma(\beta)\Gamma(\gamma)}{\Gamma(\beta + \gamma)} \int_{0}^{1} \frac{\Gamma(\beta + \gamma)}{\Gamma(\beta)\Gamma(\gamma)}z^{\beta - 1} (1 - z)^{\gamma - 1}dz$$

$$= \frac{\Gamma(\alpha + \beta + \gamma)}{\Gamma(\alpha)\Gamma(\beta + \gamma)}u^{\alpha - 1} (1 - u)^{\beta + \gamma - 1}$$

$$0 < u < 1.$$

Thus, $U \sim \text{beta}(\alpha, \beta + \gamma)$.

4.24 (345 & 445: 2 pts.) Let X and Y be independent random variables with $X \sim \text{gamma}(r, 1)$ and $Y \sim \text{gamma}(s, 1)$. Show that $Z_1 = X + Y$ and $Z_2 = X/(X + Y)$ are independent, and find the distribution of each. (Z_1 is gamma and Z_2 is beta.)

Let $z_1 = x + y$, $z_2 = \frac{x}{x+y}$, then $x = z_1 z_2$, $y = z_1 (1 - z_2)$ and

$$J = \begin{vmatrix} \frac{\partial x}{\partial z_1} & \frac{\partial x}{\partial z_2} \\ \frac{\partial y}{\partial z_1} & \frac{\partial y}{\partial z_2} \end{vmatrix} = \begin{vmatrix} z_2 & z_1 \\ 1 - z_2 & -z_1 \end{vmatrix} = z_1.$$

The set $\{x > 0, y > 0\}$ is mapped onto the set $\{z_1 > 0, 0 < z_2 < 1\}$.

$$f_{Z_1,Z_2}(z_1,z_2) = \frac{1}{\Gamma(r)} (z_1 z_2)^{r-1} e^{-z_1 z_2} \frac{1}{\Gamma(s)} (z_1 - z_1 z_2)^{s-1} e^{-z_1 + z_1 z_2} z_1$$

$$= \frac{1}{\Gamma(r+s)} z_1^{r+s-1} e^{-z_1} \frac{\Gamma(r+s)}{\Gamma(r)\Gamma(s)} z_2^{r-1} (1-z_2)^{s-1}, \qquad 0 < z_1, 0 < z_2 < 1.$$

 $f_{Z_1,Z_2}(z_1,z_2)$ can be factorized into two densities. Therefore Z_1 and Z_2 are independent and $Z_1 \sim \text{gamma } (r+s,1), Z_2 \sim \text{beta}(r,s)$.