STAT 542 Homework 8

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1. (a)

$$f(x,y) = f(y|x)f(x)$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{(y-x)^2}{2}} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

$$= \frac{1}{2\pi} e^{-\frac{(y-x)^2 + x^2}{2}}$$

(b)

$$f(y) = \int_{-\infty}^{\infty} f(x, y) dx$$

$$= \int_{-\infty}^{\infty} \frac{1}{2\pi} e^{-\frac{(y-x)^2 + x^2}{2}} dx$$

$$= \int_{-\infty}^{\infty} \frac{1}{2\pi} e^{-\frac{2x^2 - 2xy + y^2}{2}} dx$$

$$= \int_{-\infty}^{\infty} \frac{1}{2\pi} e^{-\frac{(x - \frac{1}{2}y)^2 + \frac{1}{2}y^2}{2}} dx$$

$$= \frac{\sqrt{1/2}}{\sqrt{2\pi}} e^{\frac{1}{4}y^2} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sqrt{\frac{1}{2}}} e^{-\frac{(x - \frac{1}{2}y)^2}{2(\sqrt{\frac{1}{2}})^2}} dx$$

$$= \frac{1}{2\sqrt{\pi}} e^{-\frac{1}{4}y^2}$$

Thus $Y \sim N(0,2)$

(c)

$$f(w) = \frac{1}{\sqrt{2\pi}\sqrt{2}} e^{-\frac{w^2}{2(\sqrt{2})^2}}$$

$$f(v|w) = \frac{f(v,w)}{f(w)}$$

$$= \frac{1}{2\pi} e^{\frac{(w-v)^2+v^2}{2}} / \frac{1}{2\sqrt{\pi}} e^{\frac{w^2}{4}}$$

$$= \frac{1}{\sqrt{\pi}} e^{\frac{2(w-v)^2+2v^2-w^2}{4}}$$

$$= \frac{1}{\sqrt{\pi}} e^{(v-\frac{1}{2}w)^2}$$

$$= \frac{1}{\sqrt{2\pi}\sqrt{\frac{1}{2}}} e^{\frac{(v-\frac{1}{2}w)^2}{2\cdot\frac{1}{2}}}$$

Thus $V|W=w\sim N(\frac{1}{2}w,\frac{1}{2})$

2. $D \sim unif(8, 8.5), T \sim unif(2/3, 5/6).$ $f(d, t) = f_D(d)f_T(t) = \frac{1}{8.5 - 8} \frac{1}{5/6 - 2/3} = 12.$ Thus

$$P(D+T \le 9) = \int_{2/3}^{5/6} \int_{8}^{9-t} 12 dd dt = 1/2$$

3. (a) P(X=1)=1/4, P(X=2)=1/2, P(X=3)=1/4, P(Y=2)=1/3, P(Y=3)=1/3, P(Y=4)=1/3. Thus $P(X=1)P(Y=4)=1/4\times 1/3=1/12\neq P(X=1,Y=4)=0.$ Hence, X,Y are dependent.

(b)

4. $X \sim Geometric(p), Y \sim Geometric(p).$

When $v \ge 0$, $X - Y \ge 0 \Rightarrow U = \min(X, Y) = Y$. Thus

$$P(U = u, V = v) = P(Y = u, X - Y = v)$$

$$= P(X = u + v, Y = u) = p(1 - p)^{u + v - 1}p(1 - p)^{u - 1}$$

$$= p^{2}(1 - p)^{2u + v - 2} \qquad u = 1, 2, \dots, v = 0, 1, 2, \dots$$

When v < 0, $X - Y < 0 \Rightarrow U = \min(X, Y) = X$. Thus

$$P(U = u, V = v) = P(X = u, X - Y = v)$$

$$= P(X = u, Y = u - v)$$

$$= p(1 - p)^{u-1}p(1 - p)^{u-v-1}$$

$$= p^{2}(1 - p)^{2u-v-2} \qquad u = 1, 2, \dots, v = -1, -2, \dots$$

Thus

$$f(u,v) = p^2(1-p)^{2u+|v|-2}, u = 1, 2, \dots, v = 0, \pm 1, \pm 2, \dots$$

5.

$$P(X + Y = k) = \sum_{i=1}^{k-1} P(X = i, Y = k - i)$$

$$= \sum_{i=1}^{k-1} P(X = i) P(Y = k - i)$$

$$= \sum_{i=1}^{k-1} p(1 - p)^{i-1} p(1 - p)^{k-i-1}$$

$$= \sum_{i=1}^{k-1} p^2 (1 - p)^{k-2}$$

$$= (k - 1) p^2 (1 - p)^{k-2}$$

$$P(X = x | X + Y = k) = \frac{P(X = x, X + Y = k)}{P(X + Y = k)}$$

$$= \frac{P(X = x, Y = k - x)}{P(X + Y = k)}$$

$$= \frac{p(1 - p)^{x - 1}p(1 - p)^{k - x - 1}}{(k - 1)p^{2}(1 - p)^{k - 2}}$$

$$= \frac{p^{2}(1 - p)^{k - 2}}{(k - 1)p^{2}(1 - p)^{k - 2}}$$

$$= \frac{1}{k - 1} \qquad x = 1, 2, \dots, k - 1$$

6. From

$$\begin{cases} u = x + y \\ v = x - y \end{cases}$$

we have

$$\begin{cases} x = \frac{1}{2}(u+v) \\ y = \frac{1}{2}(u-v) \end{cases}$$

Thus

$$\det J = \begin{vmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{vmatrix} = -\frac{1}{2}$$

We also have

$$f_{X,Y}(x,y) = f_X(x)f_Y(y) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(y-\gamma)^2}{2\sigma^2}} = \frac{1}{2\pi\sigma^2} e^{-\frac{(x-\mu)^2 + (y-\gamma)^2}{2\sigma^2}}$$

$$f_{U,V}(u,v) = f_{X,Y}(\frac{1}{2}(u+v), \frac{1}{2}(u-v))|\det J|$$

$$= \frac{1}{2} \frac{1}{2\pi\sigma^2} e^{-\frac{(\frac{1}{2}(u+v)-\mu)^2+(\frac{1}{2}(u-v)-\gamma)^2}{2\sigma^2}}$$

$$= \frac{1}{4\pi\sigma^2} e^{-\frac{\frac{1}{2}u^2+\frac{1}{2}v^2-\mu(u+v)-\gamma(u-v)+\mu^2+\gamma^2}{2\sigma^2}}$$

$$= \frac{1}{4\pi\sigma^2} e^{-\frac{(u-(\mu+\gamma))^2+(v-(\mu-\gamma))^2}{4\sigma^2}}$$

$$= \frac{1}{\sqrt{2\pi}\sqrt{2}\sigma} e^{-\frac{(u-(\mu+\gamma))^2}{2(\sqrt{2}\sigma)^2}} \cdot \frac{1}{\sqrt{2\pi}\sqrt{2}\sigma} e^{-\frac{(v-(\mu-\gamma))^2}{2(\sqrt{2}\sigma)^2}}$$

$$= g(u)f(v) = f_U(u) \cdot f_V(v)$$

U, V are independent, and $U \sim N(\mu + \gamma, 2\sigma^2), V \sim N(\mu - \gamma, 2\sigma^2)$

7.

$$Cov(XY,Y) = E(XY^{2}) - E(XY)E(Y)$$

$$= E(X)E(Y^{2}) - E(X)(E(Y))^{2}$$

$$= \mu_{X}(\sigma_{Y}^{2} + \mu_{Y}^{2}) - \mu_{X}\mu_{Y}^{2}$$

$$= \mu_{X}\sigma_{Y}^{2}$$

$$\begin{split} Var(XY) &= E(X^2Y^2) - (E(XY))^2 \\ &= E(X^2)E(Y^2) - (E(X)E(Y))^2 \\ &= (\sigma_X^2 + \mu_X^2)(\sigma_Y^2 + \mu_Y^2) - \mu_X^2 \mu_Y^2 \\ &= \sigma_X^2 \mu_Y^2 + \mu_X^2 \sigma_Y^2 + \sigma_X^2 \sigma_Y^2 \end{split}$$

$$Corr(X,Y) = \frac{Cov(XY,Y)}{\sqrt{Var(XY)}\sqrt{Var(Y)}}$$

$$= \frac{\mu_X \sigma_Y^2}{\sqrt{\sigma_X^2 \mu_Y^2 + \mu_X^2 \sigma_Y^2 + \sigma_X^2 \sigma_Y^2} \sigma_Y}$$

$$= \frac{\mu_X \sigma_Y}{\sqrt{\sigma_X^2 \mu_Y^2 + \mu_X^2 \sigma_Y^2 + \sigma_X^2 \sigma_Y^2}}$$

8. (a) No. With different X_1 and X_3 , according to the support of (X_1, X_2, X_3) , X_2 has different support.

(b) $f_{X_2}(x_2) = \int_0^1 \int_0^{x_2} f(x_1, x_2, x_3) dx_1 dx_3 = 6x_2(1 - x_2) = \frac{1}{B(2, 2)} x_2^{2-1} (1 - x_2)^{2-1}$

Thus $X_2 \sim Beta(2,2)$

(c) Under the condition $X_2 = x_2$, the support of (X_1, X_2) is a rectangle $(0, x_2) \times (x_2, 1)$.

$$f(x_1, x_2 | x_3) = \frac{f(x_1, x_2, x_3)}{f_{X_2}(x_2)} = \frac{6}{6x_2(1 - x_2)}$$

(d) Given $X_2 = x_2$, the support of (X_1, X_2) is a rectangle $(0, x_2) \times (x_2, 1)$, and the joint pdf can be written as $g(x_1)h(x_3)$ when we view x_2 as a constant.

$$f(x_1, x_2 | x_3) = \frac{1}{x_2} \frac{1}{1 - x_2} = f(x_1 | x_2) f(x_3 | x_2)$$

Hence X_1 and X_3 are independent conditioning on $X_2 = x_2$. And $X_1|X_2 = x_2 \sim unif(0, x_2)$, $X_3|X_2 = x_2 \sim unif(x_2, 1)$.

(e) X_1 and X_3 are independent given $X_2 = x_2$, thus

$$Cov(X_1|X_2 = x_2, X_3|X_2 = x_2) = E(X_1X_3|X_2 = x_2) - E(X_1|X_2 = x_2)E(X_3|X_2 = x_2)$$

$$= E(X_1|X_2 = x_2)E(X_3|X_2 = x_2) - E(X_1|X_2 = x_2)E(X_3|X_2 = x_2)$$

$$= 0$$

9. X and Y are independent, thus $f(x,y) = f_X(x)f_Y(y) = 1$.

$$F_Z(z) = P(Z \le z) = P(\sqrt{X+Y} \le z) = P(X+Y \le z^2) = \int_0^{z^2} \int_0^{z^2-x} dy dx = \frac{1}{2}z^4$$

Hence $f_Z(z) = \frac{d}{dz} F_Z(z) = 2z^3, z > 0.$