# MATH230: Homework 7 (due Oct. 30)

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Last compiled on: Sunday 29<sup>th</sup> October, 2023, 17:14

#### 1 Chapter 7 #4

Let  $Z_1 = X + Y$ ,  $Z_2 = Y$  and  $g(z_1, z_2) = P(Z_1 = z_1, Z_2 = z_2)$ . Then we can write

$$g(z_1, z_2) = f(z_1 - z_2, z_2) = \begin{cases} \frac{z_1 + z_2}{27} & (z_2 \in \{0, 1, 2\}, z_1 - z_2 \in \{0, 1, 2\}) \\ 0 & (\text{otherwise}) \end{cases}$$

In conclusion, we can calculate the marginal distribution of  $Z_1 = Z$ .

$$\begin{split} &P(Z=0)=g(0,0)=0\\ &P(Z=1)=g(1,0)+g(1,1)=\frac{1}{9}\\ &P(Z=2)=g(2,0)+g(2,1)+g(2,2)=\frac{1}{3}\\ &P(Z=3)=g(3,1)+g(3,2)=\frac{1}{3}\\ &P(Z=4)=g(4,2)=\frac{2}{9} \end{split}$$

## 2 Chapter 7 #5

Let  $w(y) = \exp(-y/2)$ . Then w(Y) = X holds, so we can write

$$g(y) = f(w(y))|w'(y)| = \begin{cases} \frac{\exp(-y/2)}{2} & (y > 0) \\ 0 & (y \le 0) \end{cases}$$

and g(y) is probability density function of Y. Thus, Y follows an exponential distribution with  $\lambda = 1/2$ .

## 3 Chapter 7 #7

Let  $\phi(w) = \sqrt{2w/m}$ . Then  $\phi(W) = V$  holds, so we can write

$$g(w) = f(\phi(w))|\phi'(w)| = \begin{cases} \frac{k\sqrt{2w}}{m^{3/2}} \exp(-2bw/m) & (w > 0) \\ 0 & (w \le 0) \end{cases}$$

which is a probability distribution of W.

#### 4 Chapter 7 #12

Let  $w_1(y_1, y_2) = y_1y_2, w_2(y_1, y_2) = y_1 - y_1y_2$ . Then  $X_1 = w_1(Y_1, Y_2), X_2 = w_2(Y_1, Y_2)$ , so

$$J = \det \begin{pmatrix} \frac{\partial w_1}{\partial y_1} & \frac{\partial w_1}{\partial y_2} \\ \frac{\partial w_2}{\partial y_1} & \frac{\partial w_2}{\partial y_2} \end{pmatrix} = \det \begin{pmatrix} y_2 & y_1 \\ 1 - y_2 & -y_1 \end{pmatrix} = -y_1$$

Then, we can write the joint probability density of  $Y_1$  and  $Y_2$  as

$$g(y_1, y_2) = f(w_1(y_1, y_2))f(w_2(y_1, y_2))|J|$$

$$= \begin{cases} e^{-y_1y_2}e^{-y_1+y_1y_2}|y_1| & (y_1y_2 > 0, y_1 - y_1y_2 > 0) \\ 0 & (\text{elsewhere}) \end{cases}$$

$$= \begin{cases} e^{-y_1}|y_1| & (y_1y_2 > 0, y_1 - y_1y_2 > 0) \\ 0 & (\text{elsewhere}) \end{cases}$$

The marginal distribution of  $Y_1$  can be obtained as

$$h_1(y_1) = \int_{-\infty}^{\infty} g(y_1, y_2) dy_2 = \int_{0}^{1} e^{-y_1} |y_1| dy_2 = e^{-y_1} |y_1|$$

for  $y_1 > 0$ . For  $y_1 \le 0$ ,  $g(y_1, y_2)$  is zero so  $h_1(y_1) = 0$ . Similarly, the marginal distribution of  $Y_2$  can be obtained as

$$h_2(y_2) = \int_{-\infty}^{\infty} g(y_1, y_2) dy_1 = \int_{0}^{\infty} e^{-y_1} |y_1| dy_1 = \int_{0}^{\infty} y_1 e^{-y_1} dy_1$$
$$= \left[ (-y_1 - 1)e^{-y_1} \right]_{0}^{\infty} = 1$$

for  $0 < y_2 < 1$ . Otherwise,  $g(y_1, y_2)$  is zero so  $h_2(y_2) = 0$ . Then,

$$h_1(y_1)h_2(y_2) = \begin{cases} e^{-y_1}|y_1| & (y_1 > 0, 0 < y_2 < 1) \\ 0 & \text{(elsewhere)} \end{cases}$$

As  $y_1y_2 > 0$  and  $y_1 - y_1y_2 > 0$  is equivalent to  $y_1 > 0$  and  $0 < y_2 < 1$ ,  $g(y_1, y_2) = h_1(y_1)h_2(y_2)$  holds so  $Y_1$  and  $Y_2$  are independent.

## 5 Chapter 7 #18

We can write

$$M_X(t) = \sum_{x=1}^{\infty} e^{tx} pq^{x-1} = \sum_{x=0}^{\infty} e^{t(x+1)} pq^x = pe^t \sum_{x=0}^{\infty} (qe^t)^x = \frac{pe^t}{1 - qe^t}$$

using the Taylor series expansion of geometric series. Now, we can write

$$E(X) = \frac{d}{dt}M_X(t)\Big|_{t=0} = \frac{pe^t}{(1-qe^t)^2}\Big|_{t=0} = \frac{p}{(1-q)^2} = \frac{1}{p}$$

So the mean of X is 1/p. We can also write

$$E(X^{2}) = \left(\frac{d}{dt}\right)^{2} M_{X}(t)\Big|_{t=0} = \frac{pe^{t}(1+qe^{t})}{(1-qe^{t})^{3}}\Big|_{t=0} = \frac{1+q}{p^{2}}$$

Thus, the variance of X is  $E(X^2) - (E(X))^2 = q/p^2$ .

### 6 Chapter 7 #23

#### 6.1 Solution for (a)

We can write

$$P(U=u) = \sum_{x=0}^{u} P(X=x)P(Y=u-x) = \sum_{x=0}^{u} (q^{x}p)(q^{u-x}p) = \sum_{x=0}^{u} q^{u}p^{2} = (u+1)q^{u}p^{2}$$

for  $u = 0, 1, 2, \dots$  (otherwise the probability is zero)

#### 6.2 Solution for (b)

For  $x \leq u$ , we can write

$$P(X = x, X + Y = u) = P(X = x, Y = u - x) = P(X = x)P(Y = u - x)$$
$$= (q^{x}p)(q^{u-x}p) = q^{u}p^{2}$$

Using the definition of conditional probability,

$$P(X = x | X + Y = u) = \frac{P(X = x, X + Y = u)}{P(X + Y = u)} = \frac{q^{u}p^{2}}{(u+1)q^{u}p^{2}} = \frac{1}{u+1}$$