Week 9 Study Guide (Solution)

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Joint Distributions

- 1. Discrete case:
 - (1) Exercise **6.2**: We have
 - a. The marginal p.m.f of X is

X	1	2	3
P	$\frac{1}{3}$	$\frac{1}{2}$	$\frac{1}{6}$

and the marginal p.m.f of Y is

Y	0	1	2	3
P	$\frac{1}{5}$	$\frac{1}{5}$	$\frac{1}{3}$	$\frac{4}{15}$

b. We need to find the ordered pairs (x,y) in the support of X and Y such that $x+y^2 \leq 2$.

$$\begin{split} P(X+Y^2 \leq 2) &= P(X=1,Y=0) + P(X=1,Y=1) + P(X=2,Y=0) \\ &= \frac{1}{15} + \frac{1}{15} + \frac{1}{10} \\ &= \frac{7}{30} \end{split}$$

- (2) Exercise **6.19**:
 - a. Marginal distribution of X is

$$\begin{array}{c|c|c|c} X & 0 & 1 \\ \hline P & 1/3 & 2/3 \end{array}$$

and marginal distribution of Y is

- $\frac{Y\mid 0\mid 1\mid 2}{P\mid 1/6\mid 1/3\mid 1/2}$ b. $p(z,w)=P(Z=z,W=w)=f_X(z)f_Y(w)$ for f_X and f_Y are marginal p.m.f of X and Y,
- 2. Continuous case:
 - (1) Let the random variables X, Y have joint density

$$f(x,y) = \begin{cases} 3(2-x)y & \text{if } 0 < y < 1 \text{ and } y < x < 2-y \\ 0 & \text{otherwise} \end{cases}$$

- a. Verify that it is a valid joint density function.
- b. Derive the marginal density for X.
- c. Calculate

$$P(X + Y \le 1)$$

Solution:

a. By properties of joint density functions:

$$\begin{split} \int_{-\infty}^{\infty} f(x,y) dx dy &= \int_{0}^{1} \left(\int_{y}^{2-y} (2-x) dx \right) 3y dy \\ &= \int_{0}^{1} \left(2x - \frac{1}{2} x^{2} \Big|_{x=y}^{x=2-y} \right) \\ &= \int_{0}^{1} (2-2y) \cdot 3y dy \\ &= \int_{0}^{1} 6y - 6y^{2} dy = 1. \end{split}$$

- b. To derive the marginal density for X, we use the formula $f_X(x) = \int_{-\infty}^{\infty} f(x,y) dy$ and separate our argument by cases:
 - i. When $x \in (0, 1)$,

$$f_X(x) = \int_0^x f(x, y) dy$$
$$= \int_0^x 3(2 - x) y dy$$
$$= 3(2 - x) \int_0^x y dy$$
$$= \frac{3}{2}(2 - x)x^2.$$

ii. When $x \in (1, 2)$,

$$f_X(x) = \int_0^{2-x} f(x, y) dy$$
$$= \int_0^{2-x} 3(2-x)y dy$$
$$= 3(2-x) \int_0^{2-x} y dy$$
$$= \frac{3}{2}(2-x)^3.$$

Please note that you can check the bounds for integration by drawing a graph.

c. To find the value of $P(X + Y \le 1)$, we need

$$\begin{split} P(X+Y \leq 1) &= \iint_{x+y \leq 1} f(x,y) dx dy \\ &= 3 \int_0^{1/2} \left(\int_y^{1-y} 2 - x dx \right) y dy \\ &= 3 \int_0^{1/2} \left(2x - \frac{1}{2} \Big|_{x=y}^{x=1-y} \right) y dy \\ &= 3 \int_0^{1/2} \left(\frac{3}{2} - 3y \right) y dy \\ &= \frac{3}{16} \end{split}$$

- (2) Exercise **6.5**: $f(x,y) = \frac{12}{7}(xy+y^2)$ for $0 \le x \le 1$ and $0 \le y \le 1$, and zero otherwise.
 - a. Just to check that $\iint_{\mathbb{R}^2} f(x,y) dx dy = 1$

Proof.

$$\iint_{\mathbb{R}^2} f(x,y) dx dy = \int_0^1 \int_0^1 \frac{12}{7} (xy + y^2) dx dy$$
$$= \frac{12}{7} \int_0^1 \frac{1}{2} x^2 y + xy^2 \Big|_{x=0}^{x=1} dy$$
$$= \frac{12}{7} \int_0^1 \frac{1}{2} y + y^2 dy$$
$$= \frac{12}{7} \left(\frac{1}{2} \cdot \frac{1}{2} + \frac{1}{3} \right) = 1$$

b. Marginal for X:

$$f_X(x) = \int_{\mathbb{R}} f(x,y)dy = \int_0^1 \frac{12}{7}(xy+y^2)dy = \frac{12}{7}(x/2+1/3) = \frac{6x}{7} + \frac{4}{7}$$

for $x \in (0, 1)$

Marginal for Y:

$$f_Y(y) = \int_{\mathbb{R}} f(x,y)dx = \int_0^1 \frac{12}{7}(xy+y^2)dx = \frac{12}{7}(y/2+y^2) = \frac{6y}{7} + \frac{12y^2}{7}$$

for $y \in (0,1)$

 \mathbf{c}

$$\begin{split} P(X < Y) &= \int_0^1 \int_0^y \frac{12}{7} (xy + y^2) dx dy \\ &= \frac{12}{7} \int_0^1 \frac{x^2}{2} y + xy^2 \bigg|_{x=0}^{x=y} dy \\ &= \frac{12}{7} \cdot \frac{3}{2} \int_0^1 y^3 dy \\ &= \frac{12}{7} \cdot \frac{3}{2} \cdot \frac{1}{4} = \frac{9}{14} \end{split}$$

d.

$$E[X^{2}Y] = \frac{12}{7} \int_{0}^{1} \int_{0}^{1} x^{2}y(xy + y^{2})dxdy$$

$$= \frac{12}{7} \int_{0}^{1} \int_{0}^{1} (x^{3}y^{2} + x^{2}y^{3})dxdy$$

$$= \frac{12}{7} \int_{0}^{1} \frac{1}{4}y^{2} + \frac{1}{3}y^{3}dy$$

$$= \frac{12}{7} \left(\frac{1}{4} \cdot \frac{1}{3} + \frac{1}{3} \cdot \frac{1}{4}\right) = \frac{2}{7}$$

- (3) Exercise **6.35**: $f_{X,Y}(x,y) = \frac{1}{4}(x+y)$ for $0 \le x \le y \le 2$
 - a. Check that f satisfies $\iint_{\mathbb{R}} f(x,y) dx dy = 1$.

Proof.

$$\iint_{\mathbb{R}} f(x,y) dx dy = \frac{1}{4} \int_{0}^{2} \int_{0}^{y} x + y dx dy$$

$$= \frac{1}{4} \int_{0}^{2} \frac{1}{2} x^{2} + xy \Big|_{x=0}^{x=y} dy$$

$$= \frac{1}{4} \cdot \frac{3}{2} \int_{0}^{2} y^{2} dy$$

$$= \frac{1}{4} \cdot \frac{3}{2} \cdot \frac{8}{3} = 1$$

b.

$$\begin{split} P(Y < 2X) &= \frac{1}{4} \int_0^2 \int_{y/2}^y x + y dx dy \\ &= \frac{1}{4} \int_0^2 xy + \frac{1}{2} x^2 \bigg|_{x=y/2}^{x=y} dy \\ &= \frac{1}{4} \int_0^2 y^2 + \frac{1}{2} y^2 - \frac{y^2}{2} - \frac{1}{2} (y/2)^2 dy \\ &= \frac{1}{4} \int_0^2 \frac{7}{8} y^2 dy \\ &= \frac{1}{4} \cdot \frac{7}{8} \cdot \frac{8}{3} = \frac{7}{12} \end{split}$$

c. for 0 < y < 2, the marginal density of Y is

$$f_Y(y) = \frac{1}{4} \int_0^y x + y dx = \frac{1}{4} \left(\frac{x^2}{2} + xy \right) \Big|_0^y = \frac{3}{8} y^2$$

and 0 otherwise.

3. Independence:

- (1) Suppose that Y is a random variable with mean 10 and standard deviation 2. Let X be a Bernoulli random variable with $p = \frac{1}{2}$, independent of Y. Now, consider a random variable Z = X + Y.
 - a. Let M(t) be the moment generating function (MGF) of Y, calculate the moment generating function of Z, denoted as $M_Z(t)$.
 - b. Using the result from part a, derive the mean and variance of Z.

Solution: The standard deviation for Y is 2 implies that its variance is Var(X) = 4.

a. Since X and Y are independent, we have

$$M_{X+Y}(t) = E\left[e^{t(X+Y)}\right] = E\left[e^{tX}\cdot e^{tY}\right] = E\left[e^{tX}\right]E\left[e^{tY}\right] = M_X(t)M_Y(t)$$

where $M_X(t) = \frac{1}{2}(1+e^t)$ and $M_Y(t) = M(t)$ is the MGF of X and Y, respectively. Therefore,

$$M_{X+Y}(t) = \frac{1}{2}(1+e^t)M(t).$$

b. To derive the mean and variance of X + Y by its MGF from part a, we need the result

$$M'(0) = E[Y] = 10$$

and

$$M''(0) = E[Y^2] = Var(Y) + (E[Y])^2 = 4 + 10 = 104$$

Thus, we have

$$\frac{d}{dt}M_{X+Y}(t) = \frac{1}{2} \left[e^t M(t) + (1 + e^t) M'(t) \right]$$

and

$$M_{X+Y}''(t) = \frac{1}{2} \frac{d}{dt} \left[e^t M(t) + (1+e^t) M'(t) \right] = \frac{1}{2} (e^t M(t) + 2e^t M'(t) + (1+e^t) M''(t))$$

Thus

$$E[X+Y] = M'_{X+Y}(0) = \frac{1}{2}[1 + (1+1) \cdot 10] = 10.5$$

and

$$E[(X+Y)^2] = M_{X+Y}''(0) = \frac{1}{2}[1+2\cdot 10+2\cdot 104] = 114.5$$

and thus

$$Var(X+Y) = E[(X+Y)^{2}] - (E[X+Y])^{2} = 114.5 - 10.5^{2} = 4.25$$

(2) Exercise **6.27**:
$$X_1$$
 and X_2 satisfy $P(X_1 = 1) = P(X_1 = -1) = 1/2$, $P(X_2 = 1) = p$ and $P(X_2 = -1) = 1 - p$. Also, X_1 and X_2 are independent. Let $Y = X_1X_2$.

$$P(Y = 1) = P(X_1 = 1, X_2 = 1) + P(X_1 = -1, X_2 = -1) = \frac{1}{2}(p+q) = \frac{1}{2}$$

$$P(Y = 1) = P(X_1 = 1, X_2 = -1) + P(X_1 = -1, X_2 = -1) = \frac{1}{2}(p+q) = \frac{1}{2}$$

$$P(X_2 = 1, Y = 1) = P(X_2 = 1, X_1 = 1) = \frac{1}{2}p = P(X_2 = 1)P(Y = 1)$$

$$P(X_2 = 1, Y = -1) = P(X_2 = 1, X_1 = -1) = \frac{1}{2}p = P(X_2 = 1)P(Y = -1)$$

$$P(X_2 = -1, Y = 1) = P(X_2 = -1, X_1 = -1) = \frac{1}{2}q = P(X_2 = -1)P(Y = 1)$$

$$P(X_2 = -1, Y = 1) = P(X_2 = 1 - X_1 = -1) = \frac{1}{2}q = P(X_2 = -1)P(Y = 1)$$

Based on the formulas from (c)-(f), we have X_2 and Y are independent.

(3) Exercise **6.32**: $p = \frac{7}{9}$ is the probability of one draws resulting in yellow or green balls. Note that $N \sim Geo(p)$, then the probability mass function for N is $P(N = k) = (2/9)^{k-1}(7/9)$ for $k \geq 1$. The joint distribution of (N, Y) is

$$P(N=k,Y=1) = P(k-1 \text{ white balls followed by a green ball})$$

$$= \left(\frac{2}{9}\right)^{k-1} \left(\frac{4}{9}\right)$$

and

$$P(N=k,Y=2) = P(k-1 \text{ white balls followed by a yellow ball})$$

$$= \left(\frac{2}{9}\right)^{k-1} \left(\frac{3}{9}\right)$$

By law of total probability, we have

$$P(Y = 1) = \sum_{k=1}^{\infty} P(Y = 1, N = k)$$

$$= \sum_{k=1}^{\infty} \left(\frac{2}{9}\right)^{k-1} \left(\frac{4}{9}\right)$$

$$= \frac{4}{9} \sum_{k=1}^{\infty} \left(\frac{2}{9}\right)^{k-1}$$

$$= \frac{4}{9} \cdot \frac{1}{1 - (2/9)} = \frac{4}{7}$$

Here, we use the geometric series $\sum_{k=1}^{\infty} p^{k-1} = \frac{1}{1-p}$ for any |p| < 1. Using the same argument (Law of Total Probability), we obtained $P(Y=2) = \frac{3}{7}$. Thus, we can see N and Y are independent: for any $k \ge 1$,

$$P(N=k,Y=1) = \left(\frac{2}{9}\right)^{k-1} \left(\frac{4}{9}\right) = \left(\frac{2}{9}\right)^{k-1} \left(\frac{7}{9}\right) \left(\frac{4}{7}\right) = P(N=k)P(Y=1)$$

$$P(N = k, Y = 2) = \left(\frac{2}{9}\right)^{k-1} \left(\frac{3}{9}\right) = \left(\frac{2}{9}\right)^{k-1} \left(\frac{7}{9}\right) \left(\frac{3}{7}\right) = P(N = k)P(Y = 2)$$

(4) Exercise 7.3: Let X_1 and X_2 be the change in price tomorrow and the day after tomorrow, with X_1 and X_2 being independent and their p.m.f given. Then,

$$P(X_1 + X_2 = 2) = P(X = -1, Y = 3) + P(X = 0, Y = 2) + P(X = 1, Y = 1)$$
$$+ P(X = 2, Y = 0) + P(X = 3, Y = 0)$$
$$= \frac{1}{64} + \frac{1}{64} + \frac{1}{16} + \frac{1}{64} + \frac{1}{64} = \frac{1}{8}$$

(5) Exercise **7.5**: Since X, Y, Z are three independent random variables with $X \sim \mathcal{N}(1,2), Y \sim \mathcal{N}(2,1)$, and $Z \sim \mathcal{N}(0,7)$, then W is still a normal random variable with

$$E[W] = E[X] - 4E[Y] + E[Z] = 1 - 4 \times 2 + 0 = -7$$

and

$$Var(W) = Var(X) + (-4)^{2}Var(Y) + Var(Z) = 2 + 16 + 7 = 25$$

We have

- a. $W \sim \mathcal{N}(-7, 25)$
- b. Let Φ be distribution function of standard normal random variables with $\mathcal{N}(0,1)$, we have

$$P(W > -2) = P\left(\frac{W - (-7)}{\sqrt{25}} > \frac{-2 + 7}{\sqrt{25}}\right)$$
$$= 1 - \Phi(1)$$
$$= 1 - .8413 = .1587$$

- (6) Exercise 8.9: X and Y are independent random variables with E[X] = 3, E[Y] = 5, Var(X) = 2, and Var(Y) = 3. Thus, $E[X^2] = 11$ and $E[Y^2] = 28$.
 - a. $E[3X 2Y + 7] = 3E[X] 2E[Y] + 7 = 3 \cdot 3 2 \cdot 5 + 7 = 6.$
 - b. $Var(3X 2Y + 7) = 9 \cdot Var(X) + 4 \cdot Var(Y) = 18 + 12 = 30$
 - c. We have

$$Var(XY) = E[(XY)^{2}] - (E[XY])^{2}$$
$$= E[X^{2}]E[Y^{2}] - (E[X]E[Y])^{2}$$
$$= 11 \cdot 28 - (3 \cdot 5)^{2} = 83$$

- 4. Expectation:
 - (1) Exercise 8.4: Let I_k be the indicator of the event that hat the number 4 is showing on the k-sided die. Then $Z = I_4 + I_6 + I_{12}$ with $E[I_k] = \frac{1}{k}$. Thus, by linearity of expectation:

$$E[Z] = \frac{1}{4} + \frac{1}{6} + \frac{1}{12} = \frac{1}{2}$$

(2) Exercise 8.7: Let X_B be the event that Ben calls Adam, and similar for X_C and X_D . Then, $X = X_C + X_D + X_B$ with

$$E[X] = E[X_B] + E[X_C] + E[X_D] = .3 + .4 + .7 = 1.4$$

and

$$Var(X) = Var(X_B) + Var(X_C) + Var(X_D) = (.3)(1 - .3) + (.4)(1 - .4) + (.7)(1 - .7) = .66$$

(3) Exercise 8.11: Continue our argument from exercise 8.4:

$$M_X(t) = M_{X_B}(t) \cdot M_{X_C}(t) \cdot M_{X_D}(t) = (.3e^t + .7)(.4e^t + .6)(.7e^t + .3)$$

Remark. We know that $X_B \sim \text{Ber}(.3)$, which implies that its moment generating function $M_{X_B}(t) = .3e^t + .7$, and similar argument for X_C and X_D .

- 5. Special Distributions:
 - (1) Suppose that X_1 and X_2 are independent random variables, and X_1 and X_2 have the exponential distribution with parameters β_1 and β_2 , respectively. Then,
 - a. Identify the joint density function $f(x_1, x_2)$ of X_1 and X_2 .
 - b. Use part a to show that the probability

$$P(X_1 > X_2) = \frac{\beta_2}{\beta_1 + \beta_2}$$

Solution:

a. Given that the marginal density for X_1 and X_2 are

$$f_{X_1}(x_1) = \begin{cases} \beta_1 e^{-\beta_1 x_1} & x_1 > 0\\ 0 & x_1 \le 0 \end{cases}$$

and

$$f_{X_2}(x_2) = \begin{cases} \beta_2 e^{-\beta_2 x_2} & x_2 > 0\\ 0 & x_2 \le 0 \end{cases}$$

Since X_1 and X_2 are independent, the joint density is

$$f(x_1, x_2) = f_{X_1}(x_1) \cdot f_{X_2}(x_2) \begin{cases} \beta_1 \beta_2 e^{-(\beta_1 x_1 + \beta_2 x_2)} & x_1, x_2 > 0 \\ 0 & Otherwise \end{cases}$$

b. To verify the equation above, we need to do the integration on the joint density over the area $\{x_1 > x_2\}$. In other words,

$$\begin{split} P(X_1 > X_2) &= \iint_{x_1 > x_2} f(x_1, x_2) dx_1 dx_2 \\ &= \int_0^\infty \int_{x_2}^\infty \beta_1 \beta_2 e^{-(\beta_1 x_1 + \beta_2 x_2)} dx_1 dx_2 \\ &= \beta_1 \beta_2 \int_0^\infty e^{-\beta_2 x_2} \left(\frac{1}{\beta_1} e^{-\beta_1 x_2}\right) dx_2 \\ &= \frac{\beta_2}{\beta_1 + \beta_2} \end{split}$$

Note. You can check the bounds for integration by drawing the graph.

- (2) Exercise **6.6**: In this problem, you can use the fact that $\int_0^\infty x e^{-\alpha x} dx = \frac{1}{\alpha^2}$ and $\int_0^\infty x^2 e^{-\alpha x} dx = \frac{2}{\alpha^3}$ for $\alpha > 0$.
 - i. The marginal density for X is: for x > 0

$$f_X(x) = \int_0^\infty x e^{-x(1+y)} dy = x e^{-x} \int_0^\infty e^{-xy} dy = e^{-x}$$

and zero density if $x \leq 0$. The marginal density for Y is: for y > 0,

$$f_Y(y) = \int_0^\infty x e^{-x(1+y)} dy = \frac{1}{(1+y)^2}$$

and zero density if $y \leq 0$.

ii. The expected value of XY can be obtained by

$$\begin{split} E[XY] &= \int_0^\infty \int_0^\infty xy \cdot x e^{-x(1+y)} dx dy \\ &= \int_0^\infty x^2 e^{-x} \left(\int_0^\infty y e^{-xy} dy \right) dx \\ &= \int_0^\infty x^2 e^{-x} \cdot \frac{1}{x^2} dx \\ &= \int_0^\infty e^{-x} dx = 1 \end{split}$$

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iii. The expected value of $\frac{X}{1+Y}$ can be obtained by

$$\begin{split} E\left[\frac{X}{1+Y}\right] &= \int_0^\infty \int_0^\infty \frac{x}{1+y} \cdot x e^{-x(1+y)} dx dy \\ &= \int_0^\infty \frac{1}{1+y} \left(\int_0^\infty x^2 e^{-x(1+y)} dx\right) dy \\ &= \int_0^\infty \frac{1}{1+y} \cdot \frac{2}{(1+y)^3} dy \\ &= 2 \int_0^\infty (1+y)^{-4} dy = \frac{2}{3} \end{split}$$

(3) Exercise **6.11**: Given the density of X is

$$f_X(x) = \begin{cases} 2x, & 0 < x < 1 \\ 0, & o.w. \end{cases}$$

 $Y \sim \text{Unif}[1,2]$, and X and Y are independent, the joint distribution of X and Y is

$$f(x,y) = f_X(x)f_Y(y) = \begin{cases} 2x, & 0 < x < 1, 1 < y < 2 \\ 0, & o.w. \end{cases}$$

Then, we have

$$P\left(Y - X \ge \frac{3}{2}\right) = \iint_{y - x > 3/2} f(x, y) \, dx dy$$
$$= \int_0^{1/2} \int_{x + 3/2}^2 2x \, dy \, dx$$
$$= \int_0^{1/2} 2x \left(\frac{1}{2} - x\right) \, dx$$
$$= \frac{1}{24}$$

(4) Exercise **6.12**: Note that for any $\alpha > 0$, we have $\int_0^\infty e^{-\alpha x} dx = \frac{1}{\alpha}$, thus

$$g(x) = \begin{cases} \alpha e^{-\alpha x} & x > 0\\ 0 & x \le 0 \end{cases}$$

is a valid p.d.f function. Using this argument, for x > 0 and y > 0 we have

$$f(x,y) = e^{-x} \cdot 2e^{-2y} = f_X(x)f_Y(y)$$

implies that X and Y are independent, where $f_X(x)$ and $f_Y(y)$ are two marginal density of X and Y, respectively.

Note: This study guide is used for Botao Jin's sections only. Comments, bug reports: b_jin@ucsb.edu