## Rensselaer Polytechnic Institute Department of Electrical, Computer, and Systems Engineering ECSE 2500: Engineering Probability, Spring 2023 Homework #7 Solutions

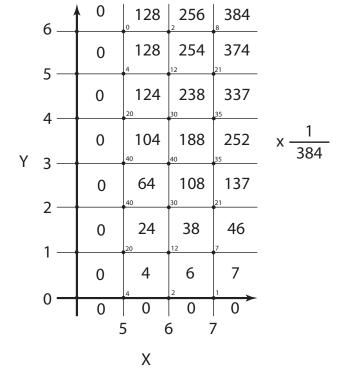
1. (a) Each value of X in [5,7] has a  $\frac{1}{3}$  probability of occurring since the distribution is uniform. If we let E be the number of eggs available, the distribution of E given X is a binomial with probability of success  $\frac{1}{2}$ . The number of eggs is the carton Y will then be  $\max(6, E)$ . That is, once we have 6 eggs, we don't collect any more. So the joint PMF table looks like:

|   |   |   | X   |   |
|---|---|---|---|---|
|   |   | 5   | 6   | 7   |
|   | 0 | $\frac{1}{3} \cdot \left(\frac{1}{2}\right)^5$  | $\frac{1}{3} \cdot \left(\frac{1}{2}\right)^6$  | $\frac{1}{3} \cdot \left(\frac{1}{2}\right)^7$  |
|   | 1 | $\frac{1}{3} \cdot 5 \cdot \left(\frac{1}{2}\right)^5$  | $\frac{1}{3} \cdot 6 \cdot \left(\frac{1}{2}\right)^6$  | $\frac{1}{3} \cdot 7 \cdot \left(\frac{1}{2}\right)^7$  |
|   | 2 | $\frac{1}{3} \cdot {5 \choose 2} \cdot {\left(\frac{1}{2}\right)}^5$  | $\frac{1}{3} \cdot {6 \choose 2} \cdot \left(\frac{1}{2}\right)^6$  | $\frac{1}{3} \cdot {7 \choose 2} \cdot \left(\frac{1}{2}\right)^7$                              |
| Y | 3 | $\frac{1}{3} \cdot {5 \choose 3} \cdot \left(\frac{1}{2}\right)^5$  | $\frac{1}{3} \cdot {6 \choose 3} \cdot \left(\frac{1}{2}\right)^6$  | $\frac{1}{3} \cdot {7 \choose 3} \cdot \left(\frac{1}{2}\right)^7$                              |
|   | 4 | $\frac{1}{3} \cdot {5 \choose 4} \cdot \left(\frac{1}{2}\right)^5$  | $\frac{1}{3} \cdot {6 \choose 4} \cdot \left(\frac{1}{2}\right)^6$  | $\frac{1}{3} \cdot {7 \choose 4} \cdot {1 \over 2}^7$   |
|   | 5 | $\frac{1}{3} \cdot {5 \choose 4} \cdot \left(\frac{1}{2}\right)^5$ $\frac{1}{3} \cdot \left(\frac{1}{2}\right)^5$ | $\frac{1}{3} \cdot {6 \choose 5} \cdot \left(\frac{1}{2}\right)^6$ $\frac{1}{3} \cdot \left(\frac{1}{2}\right)^6$ | $\frac{1}{3} \cdot {7 \choose 5} \cdot \left(\frac{1}{2}\right)^7$                              |
|   | 6 | 0   | $\frac{1}{3} \cdot \left(\frac{1}{2}\right)^6$  | $\frac{1}{3} \cdot {7 \choose 6} \cdot \left(\frac{1}{2}\right)^7 + \left(\frac{1}{2}\right)^7$ |

or getting everything into a common denominator,

|   |   |                  | X                |                  |
|---|---|------------------|------------------|------------------|
|   |   | 5                | 6                | 7                |
|   | 0 | $\frac{4}{384}$  | $\frac{2}{384}$  | $\frac{1}{384}$  |
|   | 1 | $\frac{20}{384}$ | $\frac{12}{384}$ | $\frac{7}{384}$  |
|   | 2 | $\frac{40}{384}$ | $\frac{30}{384}$ | $\frac{21}{384}$ |
| Y | 3 | $\frac{40}{384}$ | $\frac{40}{384}$ | $\frac{35}{384}$ |
|   | 4 | $\frac{20}{384}$ | $\frac{30}{384}$ | $\frac{35}{384}$ |
|   | 5 | $\frac{4}{384}$  | $\frac{12}{384}$ | $\frac{21}{384}$ |
|   | 6 | 0                | $\frac{2}{384}$  | $\frac{8}{384}$  |

(b) The joint CDF looks like a stairstep function that's easiest to represent as a grid. Remember that the joint CDF is the sum of all probability to the left and below a given point (indicated by small numbers at the dots); we accumulate probability every time we cross a line.



(c) From the joint PMF in part (a), it's easy to compute the marginal PMF of Y just by summing up the rows; we obtain

$$\begin{array}{c|c} p_{y}(Y) \\ \hline 0 & \frac{7}{384} \\ 1 & \frac{39}{384} \\ 2 & \frac{91}{384} \\ 3 & \frac{115}{384} \\ 4 & \frac{85}{384} \\ 5 & \frac{37}{384} \\ 6 & \frac{10}{384} \\ \end{array}$$

As expected these probabilities sum to 1, since the marginal is a valid PMF.

2. (a) Let's integrate this function and see what we get:

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) \, dx \, dy = k \int_{0}^{1} \int_{0}^{1} x(1 - x) y \, dx \, dy$$

$$= k \left( \frac{1}{2} x^{2} - \frac{1}{3} x^{3} \right)_{x=0}^{x=1} \right) \left( \frac{1}{2} y^{2} \right)_{y=0}^{y=1}$$

$$= k \left( \frac{1}{6} \right) \left( \frac{1}{2} \right)$$

$$= \frac{k}{12}$$

Since for a valid PDF we need this integral to equal 1, this means that k = 12.

(b) The joint CDF in the "interesting" range  $x \in [0,1]$ ,  $y \in [0,1]$  is computed as

$$F_{XY}(x,y) dx dy = \int_0^x \int_0^y f_{XY}(x,y) dx dy$$
$$= 12 \left( \frac{1}{2} x^2 - \frac{1}{3} x^3 \right) \left( \frac{1}{2} y^2 \right)$$
$$= (3x^2 - 2x^3) y^2$$

(c) To compute the marginals, we integrate out the variable we don't care about. For  $x \in [0,1]$  we have

$$f_X(x) = \int_0^1 12x(1-x)y \, dy$$

$$= (12x(1-x)) \left(\frac{1}{2}y^2\right]_{y=0}^{y=1}$$

$$= (12x(1-x))(\frac{1}{2})$$

$$= 6x(1-x) \quad x \in [0,1], 0 \text{ otherwise}$$

(d) Similarly for  $y \in [0, 1]$  we have

$$f_Y(y) = \int_0^1 12x(1-x)y \, dx$$

$$= y \left( 6x^2 - 4x^3 \right)_{x=0}^{x=1}$$

$$= y(2)$$

$$= 2y \quad y \in [0,1], 0 \text{ otherwise}$$

(e) Yes, *X* and *Y* are independent since we can see that

$$f_{XY}(x, y) = 12x(1-x)y$$
  
=  $(6x(1-x))(2y)$   
=  $f_X(x) f_Y(y)$ 

$$P(Y < \sqrt{X}) = \int_0^1 \int_0^{\sqrt{x}} 12x(1-x)y \, dy \, dx$$

$$= \int_0^1 (12x(1-x)) \left(\frac{1}{2}y^2\right]_{y=0}^{y=\sqrt{x}} dx$$

$$= \int_0^1 (12x(1-x)) \left(\frac{1}{2}x\right) \, dx$$

$$= \int_0^1 6x^2(1-x) \, dx$$

$$= 2x^3 - \frac{6}{4}x^4\Big|_{x=0}^{x=1}$$

$$= \frac{1}{2}$$

## 3. (a) We're told that *X* and *Y* are jointly Gaussian with the PDF

$$f_{XY}(x, y) = c \exp\left(-\frac{3}{64}(12x^2 - 80x + 3y^2 + 24y - 4xy)\right)$$

where *c* is an unknown constant. We are also told that the correlation coefficient  $\rho = \frac{1}{3}$ . We need to pattern-match this against the 2D Gaussian PDF of the form

$$f_{XY}(x,y) = \frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \exp\left(\frac{-1}{2(1-\rho^2)} \left[ \left(\frac{x-\mu_x}{\sigma_x}\right)^2 - 2\rho\left(\frac{x-\mu_x}{\sigma_x}\right) \left(\frac{y-\mu_y}{\sigma_y}\right) + \left(\frac{y-\mu_y}{\sigma_y}\right)^2 \right] \right)$$

We don't care about the constant in front; let's just expand the exponent, plugging in  $\rho = \frac{1}{3}$ :

$$\frac{-1}{2(1-\rho^{2})\sigma_{x}^{2}}x^{2} - \frac{1}{2(1-\rho^{2})\sigma_{y}^{2}}y^{2} + \left(\frac{\mu_{x}}{(1-\rho^{2})\sigma_{x}^{2}} - \frac{\rho\mu_{y}}{(1-\rho^{2})\sigma_{x}\sigma_{y}}\right)x + \left(\frac{\mu_{y}}{(1-\rho^{2})\sigma_{y}^{2}} - \frac{\rho\mu_{x}}{(1-\rho^{2})\sigma_{x}\sigma_{y}}\right)y + \frac{\rho}{(1-\rho^{2})\sigma_{x}\sigma_{y}}xy + \text{constant}$$

$$= \frac{-9}{16\sigma_{x}^{2}}x^{2} - \frac{9}{16\sigma_{y}^{2}}y^{2} + \left(\frac{9\mu_{x}}{8\sigma_{x}^{2}} - \frac{3\mu_{y}}{8\sigma_{x}\sigma_{y}}\right)x + \left(\frac{9\mu_{y}}{8\sigma_{y}^{2}} - \frac{3\mu_{x}}{8\sigma_{x}\sigma_{y}}\right)y + \frac{3}{8\sigma_{x}\sigma_{y}}xy + \text{constant}$$

Matching up the coefficients on the  $x^2$  and  $y^2$  terms gives us

$$\frac{-9}{16\sigma_x^2} = \frac{-9}{16} \qquad \frac{-9}{16\sigma_y^2} = \frac{-9}{64}$$

which tells us that  $\sigma_x = 1$  and  $\sigma_y = 2$ . As a sanity check we can see that the xy term also agrees.

(b) Now that we know  $\sigma_x$  and  $\sigma_y$  we can look at the x and y terms:

$$\frac{9\mu_x}{8} - \frac{3\mu_y}{16} = \frac{15}{4} \longrightarrow 18\mu_x - 3\mu_y = 60$$

$$\frac{9\mu_y}{32} - \frac{3\mu_x}{16} = \frac{-9}{8} \longrightarrow -6\mu_x + 9\mu_y = -36$$

Solving this linear system gives  $\mu_x = 3$ ,  $\mu_y = -2$ .

4. (a) The correlation of X and Y is defined as E(XY).

$$E(XY) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f_{XY}(x, y) \, dx \, dy$$

$$= \frac{6}{19} \int_{0}^{2} \int_{0}^{1} xy (x^{2} + y^{3}) \, dy \, dx = \frac{6}{19} \int_{0}^{2} \frac{1}{2} x^{3} y^{2} + \frac{1}{5} x y^{5} \Big]_{y=0}^{y=1} \, dx$$

$$= \frac{6}{19} \int_{0}^{2} \frac{1}{2} x^{3} + \frac{1}{5} x \, dx$$

$$= \frac{6}{19} \left( \frac{1}{8} x^{4} + \frac{1}{10} x^{2} \right]_{x=0}^{x=2}$$

$$= \frac{6}{19} \cdot \frac{12}{5}$$

$$= \frac{72}{95}$$

(b) The covariance of X and Y is defined as E(XY) - E(X)E(Y), which means we first have to compute the marginal means E(X) and E(Y).

$$E(X) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f_{XY}(x, y) \, dx \, dy$$

$$= \frac{6}{19} \int_{0}^{2} \int_{0}^{1} x (x^{2} + y^{3}) \, dy \, dx \qquad = \frac{6}{19} \int_{0}^{2} x^{3} y + \frac{1}{4} x y^{4} \Big|_{y=0}^{y=1} \, dx$$

$$= \frac{6}{19} \int_{0}^{2} x^{3} + \frac{1}{4} x \, dx$$

$$= \frac{6}{19} \left( \frac{1}{4} x^{4} + \frac{1}{8} x^{2} \right)_{x=0}^{x=2}$$

$$= \frac{6}{19} \cdot \frac{9}{2}$$

$$= \frac{27}{19}$$

$$E(Y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f_{XY}(x, y) \, dx \, dy$$

$$= \frac{6}{19} \int_{0}^{2} \int_{0}^{1} y (x^{2} + y^{3}) \, dy \, dx = \frac{6}{19} \int_{0}^{2} \frac{1}{2} x^{2} y^{2} + \frac{1}{5} y^{5} \Big|_{y=0}^{y=1} \, dx$$

$$= \frac{6}{19} \int_{0}^{2} \frac{1}{2} x^{2} + \frac{1}{5} \, dx$$

$$= \frac{6}{19} \left( \frac{1}{6} x^{3} + \frac{1}{5} x \right|_{x=0}^{x=2} \right)$$

$$= \frac{6}{19} \cdot \frac{26}{15}$$

$$= \frac{52}{95}$$

Thus

$$Cov(X, Y) = \frac{72}{95} - \frac{27}{19} \cdot \frac{52}{95} = \frac{-36}{1805}$$

(c) Now we use the properties of expected value and the numbers we already computed:

$$E(95(X(1+Y)+2Y(1-X))) = 95E(X+2Y-XY)$$

$$= 95(E(X)+2E(Y)-E(XY))$$

$$= 95\left(\frac{27}{19} + \frac{104}{95} - \frac{72}{95}\right)$$

$$= 167$$

(d) No, X and Y are not uncorrelated since  $E(XY) \neq 0$ .