EE306002: Probability

Spring 2022

Department of Electrical Engineering National Tsing Hua University

Homework #5 – Solutions

Coverage: Chapters 8 and 9

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Problem 8.1.3 (10 points) Let the joint probability mass function of discrete random variables X and Y be given by

$$p(x,y) = \begin{cases} k(x^2 + y^2) & \text{if } (x,y) = (1,1), (1,3), (2,3), \\ 0 & \text{otherwise.} \end{cases}$$

- (a) Determine the value of the constant k.
- (b) Determine the marginal probability mass functions of X and Y.
- (c) Find E(X) and E(Y).

Solution:

(a)
$$k[(1+1)+(1+9)+(4+9)] = 1$$
 implies that $k = 1/25$.

(b)

$$p_X(1) = p(1,1) + p(1,3) = 12/25, \ p_X(2) = p(2,3) = 13/25.$$

 $p_Y(1) = p(1,1) = 2/25, \ p_Y(3) = p(1,3) + p(2,3) = 23/25.$

Therefore,

$$p_X(x) = \begin{cases} 12/25 & \text{if } x = 1, \\ 13/25 & \text{if } x = 2. \end{cases}$$

$$p_Y(y) = \begin{cases} 2/25 & \text{if } y = 1, \\ 23/25 & \text{if } y = 3. \end{cases}$$

(c)
$$E(X) = 1 \cdot \frac{12}{25} + 2 \cdot \frac{13}{25} = \frac{38}{25}, \ E(Y) = 1 \cdot \frac{2}{25} + 3 \cdot \frac{23}{25} = \frac{71}{25}.$$

Problem 8.1.14 (10 points) Let X be the proportion of customers of an insurance company who bundle their auto and home insurance policies. Let Y be the proportion of customers who insure at least their car with the insurance company. An actuary has discovered that for, $0 \le x \le y \le 1$, the joint distribution function of X and Y is $F(x,y) = x(y^2 + xy - x^2)$. Find the expected value of the proportion of the customers of the insurance company who bundle their auto and home insurance policies.

Solution:

$$F_X(x) = P(X \le x) = P(X \le x, Y \le 1) = F(x, 1) = x(1 + x - x^2), \ 0 \le x \le 1.$$

Therefore,

$$f_X(x) = F_X'(x) = 1 + 2x - 3x^2, \ 0 \le x \le 1.$$

So

$$E(X) = \int_0^1 x(1+2x-3x^2)dx = \frac{5}{12}.$$

Problem 8.2.16 (10 points) Let X and Y be independent exponential random variables both with mean 1. Find $E[\max(X,Y)]$.

Solution:

Let F and f be the distribution and probability density functions of $\max(X, Y)$, respectively. Then

$$F(t) = P[\max(X, Y) \le t] = P(X \le t, Y \le t) = (1 - e^{-t})^2, \ t \ge 0.$$

Thus

$$f(t) = F'(t) = 2e^{-t}(1 - e^{-t}).$$

Hence

$$E[\max(X,Y)] = \int_0^\infty t \cdot 2e^{-t}(1 - e^{-t}) dt = \frac{3}{2}.$$

Problem 8.3.10 (10 points) The random variable Y is selected at random from the interval (0,1); the random variable X is then selected at random from the interval (Y,1). Find the probability density function of X.

Solution:

Let f(x,y) be the joint probability density function of X and Y. Clearly,

$$f_X(x) = \int_{-\infty}^{\infty} f_{X|Y}(x|y) f_Y(y) dy.$$

Now,

$$f_Y(y) = \begin{cases} 1 & 0 < y < 1, \\ 0 & \text{otherwise.} \end{cases}$$

and

$$f_{X|Y}(x|y) = \begin{cases} \frac{1}{1-y} & 0 < y < 1, \ y < x < 1, \\ 0 & \text{otherwise.} \end{cases}$$

Therefore, for 0 < x < 1,

$$f_X(x) = \int_0^x \frac{1}{1-y} dy = -\ln(1-x),$$

and hence

$$f_X(x) = \begin{cases} -\ln(1-x) & 0 < x < 1, \\ 0 & \text{otherwise.} \end{cases}$$

Problem 8.3.13 (10 points) The joint probability density function of X and Y is given by

$$f(x,y) = \begin{cases} ce^{-x} & \text{if } x \ge 0, \ |y| < x, \\ 0 & \text{otherwise.} \end{cases}$$

- (a) Determine the constant c.
- (b) Find $f_{X|Y}(x|y)$ and $f_{Y|X}(y|x)$.
- (c) Calculate E(Y|X=x) and Var(Y|X=x).

Solution:

(a)

$$1 = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dy dx = \int_{0}^{\infty} \int_{-x}^{x} ce^{-x} dy dx = \int_{0}^{\infty} 2cxe^{-x} dx = 2c(\left[-xe^{-x}\right]_{0}^{\infty} - \left[e^{-x}\right]_{0}^{\infty}) = 2c,$$

then

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

(b)
$$f_{X|Y}(x|y) = \frac{f(x,y)}{f_Y(y)} = \frac{\frac{1}{2}e^{-x}}{\int_{|y|}^{\infty} \frac{1}{2}e^{-x}dx} = e^{-x+|y|}, \ x > |y|.$$

$$f_{Y|X}(y|x) = \frac{f(x,y)}{f_X(x)} = \frac{\frac{1}{2}e^{-x}}{\int_{-x}^{x} \frac{1}{2}e^{-x}dy} = \frac{1}{2x}, \quad -x < y < x.$$

(c) Given X = x, Y is a uniform random variable over (-x, x) by (b). Therefore,

$$E(Y|X=x) = 0$$
, $Var(Y|X=x) = \frac{[x-(-x)]^2}{12} = \frac{x^2}{3}$.

Problem 9.1.14 (10 points) Let $X_1, X_2, ..., X_n$ be identically distributed, independent, exponential random variables with parameters $\lambda_1, \lambda_2, ..., \lambda_n$. Prove that

$$E[\min(X_1, ..., X_n)] < \min\{E(X_1), ..., E(X_n)\}.$$

Solution:

We know that

$$P(X_i \ge t) = e^{-\lambda_i t}$$
, for $i = 1, ..., n, t \ge 0$.

Therefore,

$$P[\min(X_1, X_2, ..., X_n) > t] = P(X_1 > t, X_2 > t, ..., X_n > t)$$

$$= P(X_1 > t)P(X_2 > t) \cdot \cdot \cdot P(X_n > t)$$

$$= (e^{-\lambda_1 t})(e^{-\lambda_2 t}) \cdot \cdot \cdot (e^{-\lambda_n t})$$

$$= e^{-(\lambda_1 + \lambda_2 + \dots + \lambda_n)t}.$$

which implies $\min(X_1, X_2, ..., X_n)$ is a exponential random variable with parameter $(\lambda_1 + \lambda_2 + ... + \lambda_n)$. The inequality follows from the fact that for i = 1, 2, ..., n,

$$E[\min(X_1, ..., X_n)] = \frac{1}{\lambda_1 + \dots + \lambda_n} < \frac{1}{\lambda_i} = E(X_i).$$

Problem 9.2.5 (10 points) Let $X_1, X_2, ..., X_n$ be a sequence of nonnegative, identically distributed, and independent random variables. Let F be the distribution function of X_i , $1 \le i \le n$. Prove that

$$E[X_{(n)}] = \int_0^\infty (1 - [F(x)]^n) dx.$$

Solution:

By Remark 6.4, we have

$$E[X_{(n)}] = \int_0^\infty P(X_{(n)} > x) dx.$$

Now

$$P(X_{(n)} > x) = 1 - P(X_{(n)} \le x)$$

= 1 - P(X₁ \le x, X₂ \le x, ..., X_n \le x) = 1 - [F(x)]ⁿ.

Hence

$$E[X_{(n)}] = \int_{0}^{\infty} (1 - [F(x)]^{n}) dx.$$

Problem Ch9-Review 8 (10 points) A system consists of n components whose lifetimes form an independent sequence of random variables. Suppose that the system works as long as at least one of its components works. Let $F_1, F_2, ..., F_n$ be the cumulative distribution functions (CDF) of the lifetimes of the components of the system. In terms of $F_1, F_2, ..., F_n$, find the CDF of the lifetime of the system.

Solution:

For $1 \leq i \leq n$, let X_i be the lifetime of the *i*th component. The lifetime of the system is $\max(X_1, X_2, ..., X_n)$. Let F(t) be the CDF of the lifetime of the system.

$$F(t) = P[\max(X_1, X_2, ..., X_n) \le t]$$

$$= P(X_1 \le t, X_2 \le t, ..., X_n \le t)$$

$$= P(X_1 \le t)P(X_2 \le t) \cdots P(X_3 \le t)$$

$$= F_1(t)F_2(t) \cdots F_n(t).$$

Problem 9.2.9 (10 points) Let X_1 and X_2 be two independent random variables $N(0, \sigma^2)$, and $\{X_{(1)}, X_{(2)}\}$ be the ordered statistics of $\{X_1, X_2\}$. Let $f_{12}(x_1, x_2)$ be the joint probability density function of $X_{(1)}$ and $X_{(2)}$. Find $E[X_{(1)}] = \int \int x_1 f_{12}(x_1, x_2) dx_1 dx_2$, where the integration is taken over an appropriate region.

Solution:

By Theorem 9.6,

$$f_{12}(x_1, x_2) = 2! f_{X_1}(x_1) f_{X_2}(x_2) = 2 \cdot \frac{1}{\sigma \sqrt{2\pi}} e^{-x_1^2/2\sigma^2} \cdot \frac{1}{\sigma \sqrt{2\pi}} e^{-x_2^2/2\sigma^2}$$
$$= \frac{1}{\sigma^2 \pi} e^{-x_1^2/2\sigma^2} e^{-x_2^2/2\sigma^2}, \quad -\infty < x_1 < x_2 < \infty.$$

Therefore,

$$\begin{split} E[X_{(1)}] &= \int_{-\infty}^{\infty} \int_{-\infty}^{x_2} x_1 \frac{1}{\sigma^2 \pi} e^{-x_1^2/2\sigma^2} e^{-x_2^2/2\sigma^2} dx_1 dx_2 \\ &= \frac{1}{\sigma^2 \pi} \int_{-\infty}^{\infty} e^{-x_2^2/2\sigma^2} \left(\int_{-\infty}^{x_2} x_1 e^{-x_1^2/2\sigma^2} dx_1 \right) dx_2 \\ &= \frac{1}{\sigma^2 \pi} \int_{-\infty}^{\infty} e^{-x_2^2/2\sigma^2} (-\sigma^2) e^{-x_2^2/2\sigma^2} dx_2 \\ &= -\frac{1}{\pi} \int_{-\infty}^{\infty} e^{-x_2^2/2\sigma^2} dx_2 \\ &= -\frac{1}{\pi} \cdot \sigma \sqrt{\pi} \cdot \frac{1}{\frac{\sigma}{\sqrt{2}} \sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{x_2^2}{2(\sigma/\sqrt{2})^2}} dx_2 \\ &= -\frac{1}{\pi} \cdot \sigma \sqrt{\pi} \cdot 1 = -\frac{\sigma}{\sqrt{\pi}}. \end{split}$$

Problem Ch9-Review 9 (10 points) A bar of length ℓ is broken into three pieces at two random spots. What is the probability that the length of at least one piece is less than $\ell/20$?

Solution:

Let X and Y be two random variables. X and Y are selected independently and at random

from the interval $(0, \ell)$.

$$\begin{split} &P\big[\min(X,Y-X,\ell-Y) < \frac{\ell}{20} \, \big| \, X < Y\big] P(X < Y) \\ &+ P\big[\min(Y,X-Y,\ell-X) < \frac{\ell}{20} \, \big| \, X > Y\big] P(X > Y) \\ &= 2P\big[\min(X,Y-X,\ell-Y) < \frac{\ell}{20} \, \big| \, X < Y\big] P(X < Y) \\ &= 2P\big[\min(X,Y-X,\ell-Y) < \frac{\ell}{20} \, \big| \, X < Y\big] \cdot \frac{1}{2} \\ &= 1 - P\big[\min(X,Y-X,\ell-Y) \ge \frac{\ell}{20} \, \big| \, X < Y\big] \\ &= 1 - P\big(X \ge \frac{\ell}{20}, \, Y - X \ge \frac{\ell}{20}, \, \ell - Y \ge \frac{\ell}{20} \, \big| \, X < Y\big) \\ &= 1 - P\big(X \ge \frac{\ell}{20}, \, Y - X \ge \frac{\ell}{20}, \, \ell - Y \ge \frac{19\ell}{20} \, \big| \, X < Y\big). \end{split}$$

Now $P(X \ge \frac{\ell}{20}, Y - X \ge \frac{\ell}{20}, Y \le \frac{19\ell}{20} | X < Y)$ is the area of the region

$$\{(x,y) \in R^2 \mid 0 < x < \ell, \ 0 < y < \ell, \ x \ge \frac{\ell}{20}, \ y - x \ge \frac{\ell}{20}, \ y \le \frac{19\ell}{20} \},$$

divided by the area of the region

$$\{(x.y) \in R^2 \mid 0 < x < \ell, \ 0 < y < \ell, \ y > x\}.$$

Therefore,

$$1 - P(X \ge \frac{\ell}{20}, Y - X \ge \frac{\ell}{20}, Y \le \frac{19\ell}{20} | X < Y) = 1 - \frac{\frac{1}{2} \cdot (\frac{17\ell}{20})^2}{\ell^2/2} = 1 - \frac{289}{400} = \frac{111}{400}.$$

References

[1] Saeed Ghahramani, Fundamentals of Probability: With Stochastic Processes, Chapman and Hall/CRC; 4th edition (September 4, 2018)