

Solutions - Practice problems for Exam 2
Math 464 - Fall 18

1. Let X and Y be independent random variables. They both have a gamma distribution with mean 3 and variance 3.

(a) Find the joint probability density function (pdf) of X, Y .

Solution: Since they are independent it is just the product of a gamma density for X and a gamma density for Y . For the gamma distribution, $\mu = w/\lambda$, $\sigma^2 = w/\lambda^2$. Since the mean and variance are both 3, $\lambda = 1$ and $w = 3$. So

$$f_{X,Y}(x, y) = \begin{cases} \frac{1}{\Gamma(3)^2} x^2 y^2 e^{-x-y} & \text{if } x \geq 0, y \geq 0 \\ 0, & \text{otherwise} \end{cases}$$

(b) Express $P(3X + Y \leq 3)$ as an integral. Do not try to do the integral.

Solution: The region where $3x + y \leq 3, x \geq 0, y \geq 0$ is the triangle in the upper right quadrant below the line $y \leq 3 - 3x$. So we get

$$\int_0^1 \left[\int_0^{3-3x} \frac{1}{\Gamma(3)^2} x^2 y^2 e^{-x-y} dy \right] dx$$

2. Let X have an exponential distribution with $E[X] = 1$. Let $Y = X^2 - 2$.

(a) Find the mean and variance of Y .

Solution: First we compute some moments of X for later use. The mgf for X is $m(t) = 1/(1-t)$.

$$\begin{aligned} m'(t) &= \frac{1}{(1-t)^2}, & E[X] &= m'(0) = 1, \\ m^{(2)}(t) &= \frac{2}{(1-t)^3}, & E[X^2] &= m^{(2)}(0) = 2, \\ m^{(3)}(t) &= \frac{6}{(1-t)^4}, & E[X^3] &= m^{(3)}(0) = 6, \\ m^{(4)}(t) &= \frac{24}{(1-t)^5}, & E[X^4] &= m^{(4)}(0) = 24 \end{aligned}$$

Now

$$E[Y] = E[X^2] - 2 = 2 - 2 = 0$$

$$E[Y^2] = E[(X^2 - 2)^2] = E[X^4 - 4X^2 + 4] = 24 - 4 \cdot 2 + 4 = 20$$

So $\text{var}(Y) = 20 - 0 = 20$.

(b) Find the probability density function (pdf) for Y .

Solution: We start by finding the cdf for Y .

$$\begin{aligned} F_Y(y) &= P(Y \leq y) = P(X^2 - 2 \leq y) = P(X^2 \leq y + 2) \\ &= P(X \leq \sqrt{y + 2}) = \int_0^{\sqrt{y+2}} e^{-x} dx = 1 - \exp(-\sqrt{y + 2}) \end{aligned}$$

Take the derivative of this to get

$$f_Y(y) = \frac{1}{2}(y + 2)^{-1/2} \exp(-\sqrt{y + 2}), y \geq -2$$

The range for Y is $[-2, \infty)$.

3. Let X and Y be continuous random variables with joint pdf

$$f_{X,Y}(x, y) = \frac{3}{2}(x^2 + y^2), \quad 0 \leq x \leq 1, 0 \leq y \leq 1$$

Outside of $0 \leq x \leq 1, 0 \leq y \leq 1$, $f_{X,Y}(x, y) = 0$.

(a) Find the marginal densities of X and Y . The marginal density of X for $0 \leq x \leq 1$ is

$$f_X(x) = \int_0^1 \frac{3}{2}(x^2 + y^2) dy = \frac{3}{2}[x^2 + \int_0^1 y^2 dy] = \frac{3}{2}[x^2 + \frac{1}{3}] = \frac{1}{2} + \frac{3}{2}x^2$$

So

$$f_X(x) = \begin{cases} \frac{1}{2} + \frac{3}{2}x^2 & \text{if } 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

The same calculation shows

$$f_Y(y) = \begin{cases} \frac{1}{2} + \frac{3}{2}y^2 & \text{if } 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

(b) Are X and Y independent?

Solution: They are not independent since $f_{X,Y}(x, y)$ is not equal to $f_X(x)f_Y(y)$.

4. Let X, Y be jointly continuous random variables with joint probability density function (pdf)

$$f_{X,Y}(x, y) = \begin{cases} 4xy, & \text{if } 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

Let $Z = X + Y$. Compute $f_Z(z)$, the probability density function (pdf) for Z .

Solution: The range of X, Y is the unit square. The range of Z will be $[0, 2]$ We need to compute the cdf, $P(Z \leq z) = P(X + Y \leq z)$. How the line $x + y = z$ intersects the unit square depends on whether $0 \leq z \leq 1$ or $1 \leq z \leq 2$. In the first case

$$P(X + Y \leq z) = \int_0^z \int_0^{z-x} 4xy \, dy \, dx$$

After some calculation this equals $\frac{1}{6}z^4$. For $1 \leq z \leq 2$,

$$P(X + Y \leq z) = \int_0^{z-1} \int_0^1 4xy \, dy \, dx + \int_{z-1}^1 \int_0^{z-x} 4xy \, dy \, dx$$

After an unreasonable amount of calculation this equals $-\frac{1}{6}z^4 + 2z^2 - \frac{8}{3}z + 1$. So the pdf is

$$f_Z(z) = \begin{cases} \frac{2}{3}z^3, & \text{if } 0 \leq z \leq 1 \\ -\frac{2}{3}z^3 + 4z - \frac{8}{3}, & \text{if } 1 \leq z \leq 2 \\ 0, & \text{otherwise} \end{cases}$$

5. Let X and Y be independent random variables, each of which has a standard normal pdf. Let $Z = Y - X + 4$.

(a) Find the mean and variance of Z .

Solution: $E[Z] = E[Y] - E[X] + 4 = 4$. $\text{var}(Z) = \text{var}(Y) + \text{var}(-X) = \text{var}(Y) + \text{var}(X) = 1 + 1 = 2$.

(b) Find the probability density function (pdf) of Z . Hint: this can be done with very little computation.

Solution: It is easy to show that $-X$ is also a standard normal. The sum of independent normal random variables is normal, and adding a constant to a normal random variable gives another normal random variable. So Z is

normal. Part (a) tells us its mean and variance. Another way to see this is to look at the mgf. Since X and Y are independent, functions of them are independent. So

$$\begin{aligned} M_Z(t) &= E[\exp(t(Y - X + 4))] = e^{4t} E[\exp(tY)] E[\exp(-tX)] \\ &= \exp(4t + \frac{1}{2}t^2 + \frac{1}{2}(-t)^2) = \exp(4t + t^2) \end{aligned}$$

which is mgf of a normal with mean 4 and variance 2. So

$$f_Z(z) = \frac{1}{\sqrt{4\pi}} \exp\left(-\frac{1}{4}(z - 4)^2\right), \quad -\infty < z < \infty$$

6. Random variables X and Y have joint cumulative distribution function (cdf)

$$F_{X,Y}(x, y) = \begin{cases} [\frac{1}{\pi} \tan^{-1}(x) + c](1 - e^{-y}), & \text{if } y \geq 0 \\ 0, & \text{if } y < 0 \end{cases}$$

where c is some constant.

(a) Are X and Y independent?

Solution: Yes, the joint cdf factors into a function of x times a function of y , so they are independent.

(b) Find the value of c .

Solution:

$$\lim_{x,y \rightarrow \infty} F(x, y) = \frac{1}{\pi} \frac{\pi}{2} + c = \frac{1}{2} + c$$

This must equal 1, so $c = 1/2$.

(c) Find the joint probability density function (pdf) for X, Y .

Solution: We take the second order partial derivative of $F_{X,Y}(x, y)$ with respect to x and y . This gives

$$f_{X,Y}(x, y) = \begin{cases} \frac{1}{\pi} \frac{1}{1+x^2} e^{-y}, & \text{if } y \geq 0 \\ 0, & \text{if } y < 0 \end{cases}$$

Note that X, Y are independent. X has the Cauchy distribution, and Y is exponential with $\lambda = 1$.

7. A RV X has a Laplace distribution if its pdf is

$$f_X(x) = \frac{1}{2}\lambda e^{-\lambda|x|}, \quad -\infty < x < \infty$$

where $\lambda > 0$ is a parameter.

(a) Compute the moment generating function of X .

Solution:

$$\begin{aligned} M_X(t) &= \frac{\lambda}{2} \int_{-\infty}^{\infty} e^{tx} e^{-\lambda|x|} dx \\ &= \frac{\lambda}{2} \int_{-\infty}^0 e^{tx+\lambda x} dx + \frac{\lambda}{2} \int_0^{\infty} e^{tx-\lambda x} dx \\ &= \frac{\lambda}{2} \frac{1}{t+\lambda} e^{tx+\lambda x} \Big|_{x=-\infty}^{x=0} + \frac{\lambda}{2} \frac{1}{t-\lambda} e^{tx-\lambda x} \Big|_{x=0}^{x=\infty} \\ &= \frac{\lambda}{2} \frac{1}{t+\lambda} - \frac{\lambda}{2} \frac{1}{t-\lambda} \\ &= \frac{\lambda^2}{\lambda^2 - t^2} \end{aligned}$$

(b) Find the mean and variance of X .

Solution:

$$M'(t) = \frac{2t\lambda^2}{(\lambda^2 - t^2)^2}$$

So $M'(0) = 0$. So the mean is zero.

$$M''(t) = \frac{2\lambda^2}{(\lambda^2 - t^2)^2} + \frac{8t^2\lambda^2}{(\lambda^2 - t^2)^4}$$

So $M''(0) = 2/\lambda^2$. Since the mean is zero, the variance is $2/\lambda^2$.

8. Let X_j be a sequence of independent, identically distributed random variables. Their common density is

$$f(x) = 4xe^{-2x}, x \geq 0$$

(It is zero for $x < 0$.) Let

$$\bar{X}_n = \frac{1}{n} \sum_{j=1}^n X_j$$

(a) Find the mean and variance of \bar{X}_n . Hint: this density is in our catalog.

Solution: The X_j have a gamma distribution with $w = 2$ and $\lambda = 2$. So $E[X_j] = 1$ and $\text{var}(X_j) = 1/2$. So the mean of \bar{X}_n is 1 and its variance is $\frac{1}{2n}$.

(b) For $n = 1000$, the probability that \bar{X}_{1000} is in $[1, 1.1]$ is approximately given by

$$\int_a^b \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$$

Find a and b .

Solution:

$$\begin{aligned} P(1 \leq \bar{X}_{1000} \leq 1.1) &= P\left(0 \leq \frac{\bar{X}_{1000} - 1}{\frac{1}{\sqrt{2n}}} \leq \frac{1.1 - 1}{\frac{1}{\sqrt{2n}}}\right) \\ &= P(0 \leq Z \leq 0.1\sqrt{2n}) \end{aligned}$$

So $a = 0$ and $b = 0.1\sqrt{2n} = 2\sqrt{5}$.

9. A random number generator produces random real numbers that are uniformly distributed between -1 and 1 . I call it n times and let \bar{X} be the average of the n numbers I get. (So \bar{X} is the sum of the n numbers divided by n .) As n goes to ∞ , $P(|\bar{X}| \geq 0.01)$ goes to zero. Find n so that this probability is approximately 0.05.

For a standard normal RV Z , we have $P(Z \leq -2.576) = 0.005$, $P(Z \leq -2.326) = 0.01$, $P(Z \leq -1.960) = 0.025$, $P(Z \leq -1.649) = 0.05$.

Solution: The mean of X_i is 0 and the variance is $1/3$. So the mean of \bar{X} is 0 and the variance is $\frac{1}{3n}$. So

$$Z = \frac{\bar{X} - 0}{\sqrt{\frac{1}{3n}}} = \sqrt{3n} \bar{X}$$

So we want $P(|Z| \geq \sqrt{3n} 0.01) = 0.05$. So $P(Z \leq -\sqrt{3n} 0.01) = 0.025$. So $\sqrt{3n} 0.01 = 1.96$. And so $n = (196)^2/3 = 12,805$