

Problem Set 3
Data Analysis and Interpretation (EE 223)
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1. The joint density function of X and Y is given by

$$f(x, y) = \begin{cases} ce^{-x}e^{-2y} & 0 < x < \infty, 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

Compute (a) c , (b) $P\{X > 1, Y < 1\}$, (c) $P\{X < Y\}$, and (d) $P\{X < a\}$.

(a) By definition of pdf

$$\begin{aligned} \int_0^\infty \int_0^\infty ce^{-x}e^{-2y} dx dy &= 1 \\ c \int_0^\infty e^{-x} dx \int_0^\infty e^{-2y} dy &= 1 \\ c \cdot 1 \cdot \frac{1}{2} &= 1 \\ c &= 2 \end{aligned}$$

(b)

$$\begin{aligned} P\{X > 1, Y < 1\} &= \int_0^1 \int_1^\infty 2e^{-x}e^{-2y} dx dy = - \int_0^1 2e^{-2y}(e^{-x}|_1^\infty) dy \\ &= e^{-1} \int_0^1 2e^{-2y} dy = e^{-1}(1 - e^{-2}) \end{aligned}$$

(c)

$$\begin{aligned} P\{X < Y\} &= \iint_{(x,y): x < y} 2e^{-x}e^{-2y} dx dy = \int_0^\infty \int_0^y 2e^{-x}e^{-2y} dx dy \\ &= \int_0^\infty 2e^{-2y}(1 - e^{-y}) dy = \int_0^\infty 2e^{-2y} dy - \int_0^\infty 2e^{-3y} dy \\ &= 1 - \frac{2}{3} = \frac{1}{3} \end{aligned}$$

(d)

$$P\{X < a\} = \int_0^a \int_0^\infty 2e^{-x}e^{-2y} dy dx = \int_0^a e^{-x} dx = 1 - e^{-a}$$

2. The joint density of X and Y is given by

$$f(x, y) = \begin{cases} e^{-(x+y)} & 0 < x < \infty, 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

Find the density function of the random variable X/Y .

We start by computing the distribution function of X/Y . For $a > 0$,

$$\begin{aligned} F_{X/Y}(a) &= P\left\{\frac{X}{Y} \leq a\right\} = \iint_{x/y \leq a} e^{-(x+y)} dx dy \\ &= \int_0^\infty \int_0^{ay} e^{-(x+y)} dx dy = \int_0^\infty (1 - e^{-ay})e^{-y} dy \\ &= \left\{-e^{-y} + \frac{e^{-(1+a)y}}{1+a}\right\}_0^\infty = 1 - \frac{1}{1+a} \end{aligned}$$

Differentiation shows that the density function of X/Y is given by $f_{X/Y}(a) = \frac{1}{(1+a)^2}$, $0 < a < \infty$.

3. Sonia and Narendra decide to meet at a certain location. If each of them independently arrives at a time uniformly distributed between 5 PM and 6 PM, find the probability that the first to arrive has to wait longer than 10 minutes. If we let X and Y denote, respectively, the time past 5 PM that Sonia and Narendra arrive, then X and Y are independent random variables, each of which is uniformly distributed over $(0, 60)$. The desired probability, $P\{X + 10 < Y\} + P\{Y + 10 < X\}$, which, by symmetry, equals $2P\{X + 10 < Y\}$, is obtained as follows:

$$\begin{aligned} 2P\{X + 10 < Y\} &= 2 \iint_{x+10 < y} f(x, y) dx dy = 2 \iint_{x+10 < y} f(x) f(y) dx dy \\ &= 2 \int_{10}^{60} \int_0^{y-10} \left(\frac{1}{60}\right)^2 dx dy = \frac{2}{(60)^2} \int_{10}^{60} (y-10) dy \\ &= \frac{25}{36} \end{aligned}$$

4. If the joint density function of X and Y is

$$f(x, y) = 6e^{-2x}e^{-3y} \quad 0 < x < \infty, \quad 0 < y < \infty$$

and is equal to 0 outside this region, are the random variables independent? What if the joint density function is

$$f(x, y) = 24xy \quad 0 < x < 1, \quad 0 < y < 1, \quad 0 < x + y < 1$$

and is equal to 0 otherwise?

In the first instance, the joint density function factors, and thus the random variables, are independent (with one being exponential with rate 2 and the other exponential with rate 3). In the second instance, because the region in which the joint density is nonzero cannot be expressed in the form $x \in A, y \in B$, the joint density does not factor, so the random variables are not independent. This can be seen clearly by letting

$$I(x, y) = \begin{cases} 1 & 0 < x < 1, \quad 0 < y < 1, \quad 0 < x + y < 1 \\ 0 & \text{otherwise} \end{cases}$$

and writing $f(x, y) = 24xyI(x, y)$ which clearly does not factor into a part depending only on x and another depending only on y . Alternatively, this can be done by finding $f(x)$ and $f(y)$ and evaluating whether $f(x, y) = f(x)f(y)$ or not.

5. Nitish and Lalu shoot at a target. The distance of each shot from the center of the target is uniformly distributed on $(0, 1)$, independently of the other shot. What is the PDF of the distance of the losing shot from the center? Let X and Y be the distances from the center of the first and second shots, respectively. Let also Z be the distance of the losing shot:

$$Z = \max\{X, Y\}.$$

We know that X and Y are uniformly distributed over $[0, 1]$, so that for all $z \in [0, 1]$, we have

$$P\{X \leq z\} = P\{Y \leq z\} = z.$$

Thus, using the independence of X and Y , we have for all $z \in [0, 1]$,

$$\begin{aligned} F_Z(z) &= P\{\max\{X, Y\} \leq z\} = P\{X \leq z, Y \leq z\} \\ &= P\{X \leq z\}P\{Y \leq z\} = z^2. \end{aligned}$$

Differentiating, we obtain

$$f_Z(z) = \begin{cases} 2z & 0 \leq z \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$

6. Let X, Y, Z be independent and uniformly distributed over $(0, 1)$. Compute $P\{X \geq YZ\}$.

Since $f(x, y, z) = f(x)f(y)f(z) = 1$ $0 \leq x, y, z \leq 1$, we have

$$\begin{aligned} P\{X \geq YZ\} &= \iiint_{x \geq yz} f(x, y, z) dx dy dz = \int_0^1 \int_0^1 \int_{yz}^1 dx dy dz \\ &= \int_0^1 \int_0^1 (1 - yz) dy dz = \int_0^1 \left(1 - \frac{z}{2}\right) dz \\ &= \frac{3}{4}. \end{aligned}$$

7. Sum of two independent random variables

- (a) If X and Y are independent random variables, both uniformly distributed on $(0, 1)$, calculate the probability density of $X + Y$.

We use the result that the pdf f_{X+Y} is the convolution of the pdfs f_X and f_Y .

$$f_{X+Y}(a) = \int_{-\infty}^{\infty} f_X(a - y)f_Y(y)dy = \int_0^1 f_X(a - y)dy$$

For $0 \leq a \leq 1$, this yields

$$f_{X+Y}(a) = \int_0^a dy = a$$

For $1 < a < 2$, this yields

$$f_{X+Y}(a) = \int_{a-1}^1 dy = 2 - a.$$

Because of the shape of its density function, the random variable $X + Y$ is said to have a triangular distribution

- (b) If X and Y are independent Gamma random variables with respective parameters (α_1, β) and (α_2, β) , then prove that $X + Y$ is also a Gamma random variable with parameters $(\alpha_1 + \alpha_2, \beta)$. Recall that a Gamma random variable has a density of the form

$$f(x; \alpha, \beta) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x} \quad x, \alpha, \beta > 0.$$

We use the result that the pdf f_{X+Y} is the convolution of the pdfs f_X and f_Y .

$$\begin{aligned} f_{X+Y}(a) &= \frac{1}{\Gamma(\alpha_1)\Gamma(\alpha_2)} \int_0^a \beta^{\alpha_1} (a - y)^{\alpha_1-1} e^{-\beta(a-y)} \beta^{\alpha_2} y^{\alpha_2-1} e^{-\beta y} dy \\ &= K e^{-\beta a} \int_0^a (a - y)^{\alpha_1-1} y^{\alpha_2-1} dy \\ &= K e^{-\beta a} a^{\alpha_1+\alpha_2-1} \int_0^1 (1 - z)^{\alpha_1-1} z^{\alpha_2-1} dz \quad \text{by letting } x = \frac{y}{a} \\ &= C e^{-\beta a} a^{\alpha_1+\alpha_2-1} \end{aligned}$$

where C is a constant that does not depend on a . But, as the preceding is a density function and thus must integrate to 1, the value of C is determined, and we have

$$f(x; \alpha_1 + \alpha_2, \beta) = \frac{\beta^{\alpha_1+\alpha_2}}{\Gamma(\alpha_1 + \alpha_2)} x^{\alpha_1+\alpha_2-1} e^{-\beta x}$$

Hence, the result is proved.

- (c) If X and Y are independent Gaussian random variables with respective parameters (μ_1, σ_1^2) and (μ_2, σ_2^2) , then prove that $X + Y$ is also a Gaussian random variable with parameters $(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$. Recall that a Gaussian random variable has a density of the form

$$f(x; \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}.$$

Let, $Z = X + Y$. Therefore, we have,

$$\begin{aligned} f_Z(z) &= \int_{-\infty}^{\infty} f_X(x) f_Y(z-x) dx \\ &= \frac{1}{2\pi\sigma_1\sigma_2} \int_{-\infty}^{\infty} e^{-\frac{(x-\mu_1)^2}{2\sigma_1^2}} e^{-\frac{(z-x-\mu_2)^2}{2\sigma_2^2}} dx \end{aligned}$$

Taking $a = x - \mu_1$ and $b = z - \mu_1 - \mu_2$, expressing the argument of the exponent in terms of a, b we get

$$\begin{aligned} f_Z(z) &= \frac{1}{2\pi\sigma_1\sigma_2} e^{-\frac{b^2}{2(\sigma_1^2 + \sigma_2^2)}} \left(\frac{2\pi\sigma_1^2\sigma_2^2}{\sigma_1^2 + \sigma_2^2} \right)^{1/2} \\ &= \frac{1}{\sqrt{2\pi(\sigma_1^2 + \sigma_2^2)}} e^{-\frac{(z-\mu_1-\mu_2)^2}{2(\sigma_1^2 + \sigma_2^2)}} \end{aligned}$$

This completes the proof.

8. Conditional distribution of random variable

- (a) Suppose that the joint density of X and Y is given by

$$f(x, y) = \begin{cases} \frac{e^{-x/y} e^{-y}}{y} & 0 < x < \infty, 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

Find $P\{X > 1 | Y = y\}$.

We first obtain the conditional density of X given that $Y = y$.

$$\begin{aligned} f_{X|Y}(x|y) &= \frac{f_{X,Y}(x, y)}{f_Y(y)} \\ &= \frac{e^{-x/y} e^{-y}/y}{\int_0^{\infty} e^{-x/y} e^{-y}/y dx} = \frac{e^{-x/y}}{y} \end{aligned}$$

Hence

$$\begin{aligned} P\{X > 1 | Y = y\} &= \int_1^{\infty} \frac{e^{-x/y}}{y} dx = -e^{-x/y} \Big|_1^{\infty} \\ &= e^{-1/y}. \end{aligned}$$

- (b) If X and Y are independent Poisson random variables with respective parameters λ_1 and λ_2 . Prove that the conditional distribution of X given that $X + Y = n$ is a binomial distribution. A Poisson random variable has a pmf as

$$f(k; \lambda) = P\{X = k\} = \frac{\lambda^k e^{-\lambda}}{k!}.$$

We calculate the conditional probability mass function of X given that $X + Y = n$ as follows:

$$\begin{aligned} P\{X = k | X + Y = n\} &= \frac{P\{X = k, X + Y = n\}}{P\{X + Y = n\}} = \frac{P\{X = k, Y = n - k\}}{P\{X + Y = n\}} \\ &= \frac{P\{X = k\} P\{Y = n - k\}}{P\{X + Y = n\}} \end{aligned}$$

where the last equality follows from the assumed independence of X and Y . It can be easily proven that $X + Y$ has a Poisson distribution with parameter $\lambda_1 + \lambda_2$, we see that the preceding equals

$$\begin{aligned} P\{X = k | X + Y = n\} &= \frac{\lambda_1^k e^{-\lambda_1}}{k!} \frac{\lambda_2^{n-k} e^{-\lambda_2}}{(n-k)!} \left[\frac{(\lambda_1 + \lambda_2)^n e^{-(\lambda_1 + \lambda_2)}}{n!} \right]^{-1} \\ &= \frac{n!}{(n-k)!k!} \frac{\lambda_1^k \lambda_2^{n-k}}{(\lambda_1 + \lambda_2)^n} = \binom{n}{k} \left(\frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^k \left(\frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{n-k} \end{aligned}$$

In other words, the conditional distribution of X given that $X + Y = n$ is the binomial distribution with parameters n and $\frac{\lambda_1}{\lambda_1 + \lambda_2}$.

- (c) Rajan goes to the bank to make a deposit, and is equally likely to find 0 or 1 customer ahead of him. The times of service of these customers are independent and exponentially distributed with parameter λ . What is the CDF of Rajan's waiting time? Recall that a Exponential random variable has a density of the form

$$f(x; \lambda) = \lambda e^{-\lambda x} \quad x \geq 0.$$

Let X be the waiting time and Y be the number of customers found. For $x < 0$, we have $F_X(x) = 0$. For $x \geq 0$,

$$F_X(x) = P\{X \leq x\} = \frac{1}{2}(P\{X \leq x | Y = 0\} + P\{X \leq x | Y = 1\})$$

We have

$$\begin{aligned} P\{X \leq x | Y = 0\} &= 1 \\ P\{X \leq x | Y = 1\} &= 1 - e^{-\lambda x} \end{aligned}$$

Thus

$$F_X(x) = \frac{1}{2}(2 - e^{-\lambda x}) \quad x \geq 0.$$

9. Joint probability distribution of functions of random variables

- (a) X and Y have joint density function

$$f(x, y) = \frac{1}{x^2 y^2} \quad x \geq 1, y \geq 1$$

Compute the joint density function of $U = XY$, $V = X/Y$. What are the marginal densities?

If $u = xy, v = x/y$, then $J = \begin{vmatrix} y & x \\ \frac{1}{y} & -\frac{x}{y} \end{vmatrix} = \frac{-2x}{y}$, and $x = \sqrt{uv}, y = \sqrt{u/v}$. Hence

$$f_{U,V}(u, v) = \frac{1}{2v} f_{X,Y}(\sqrt{uv}, \sqrt{u/v}) = \frac{1}{2u^2 v}, \quad u \geq 1, \frac{1}{u} < v < u$$

To evaluate marginal densities,

$$f_U(u) = \int_{1/u}^u \frac{1}{2u^2 v} dv = \frac{1}{u^2} \log u, \quad u \geq 1.$$

For $v > 1$

$$f_V(v) = \int_v^\infty \frac{1}{2u^2 v} du = \frac{1}{2v^2}, \quad v > 1$$

For $v < 1$

$$f_V(v) = \int_{1/2}^\infty \frac{1}{2u^2 v} du = \frac{1}{2}, \quad 0 < v < 1.$$

- (b) Let X be exponentially distributed with mean 1. Once we observe the experimental value x of X , we generate a normal random variable Y with zero mean and variance $x + 1$. What is the joint PDF of X and Y ? We have $f_X(x) = e^{-x}$, for $x \geq 0$, and

$$f_{Y|X}(y|x) = \frac{1}{\sqrt{2\pi(x+1)}} e^{-y^2/2(x+1)}.$$

Thus

$$f_{X,Y}(x,y) = f_X(x)f_{Y|X}(y|x) = e^{-x} \frac{1}{\sqrt{2\pi(x+1)}} e^{-y^2/2(x+1)},$$

for all $x \geq 0$ and all y .