

Mathematical Statistics - Assignment 8

Philip Warton

November 23, 2020

Problem 5.9

Let Y_1 and Y_2 have the joint probability density function given by

$$f(y_1, y_2) = \begin{cases} k(1 - y_2) & 0 \leq y_1 \leq y_2 \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

(a)

Find the value of k that makes this a probability density function.

We want to find some value of k such that the double integral on \mathbb{R}^2 evaluates to 1. We write:

$$\begin{aligned} 1 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(y_1, y_2) dy_1 dy_2 \\ &= \int_0^1 \int_0^{y_2} k(1 - y_2) dy_1 dy_2 \\ &= \int_0^1 [ky_1(1 - y_2)]_0^{y_2} dy_2 \\ &= \int_0^1 ky_2 - ky_2^2 dy_2 \\ &= \left[\frac{ky_2^2}{2} - \frac{ky_2^3}{3} \right]_0^1 \\ &= \frac{k}{2} - \frac{k}{3} \\ \implies 6 &= 3k - 2k \\ &= k \end{aligned}$$

(b)

Find $P(Y_1 \leq \frac{3}{4}, Y_2 \geq \frac{1}{2})$.

We can rewrite this as $P(Y_2 \geq \frac{1}{2}) - P(Y_1 > \frac{3}{4})$. Then this becomes an integral that we can compute:

$$\begin{aligned}
P(Y_2 \geq \frac{1}{2}) - P(Y_1 > \frac{3}{4}) &= \int_{\frac{1}{2}}^1 \int_0^{y_2} f(y_1, y_2) dy_1 dy_2 - \int_{\frac{3}{4}}^1 \int_{y_1}^1 f(y_1, y_2) dy_2 dy_1 \\
&= \int_{\frac{1}{2}}^1 \int_0^{y_2} 6(1 - y_2) dy_1 dy_2 - \int_{\frac{3}{4}}^1 \int_{y_1}^1 6(1 - y_2) dy_2 dy_1 \\
&= \int_{\frac{1}{2}}^1 [6y_1(1 - y_2)]_0^{y_2} dy_2 - \int_{\frac{3}{4}}^1 \int_{y_1}^1 6(1 - y_2) dy_2 dy_1 \\
&= \int_{\frac{1}{2}}^1 6y_2 - 6y_2^2 dy_2 - \int_{\frac{3}{4}}^1 \int_{y_1}^1 6(1 - y_2) dy_2 dy_1 \\
&= [3y_2^2 - 2y_2^3]_{\frac{1}{2}}^1 - \int_{\frac{3}{4}}^1 \int_{y_1}^1 6(1 - y_2) dy_2 dy_1 \\
&= [3 - 2] - [\frac{3}{4} - \frac{1}{4}] - \int_{\frac{3}{4}}^1 \int_{y_1}^1 6(1 - y_2) dy_2 dy_1 \\
&= \frac{1}{2} - \int_{\frac{3}{4}}^1 \int_{y_1}^1 6(1 - y_2) dy_2 dy_1 \\
&= \frac{1}{2} - \int_{\frac{3}{4}}^1 6y_2 - 3y_2^2 \Big|_{y_1}^1 dy_1 \\
&= \frac{1}{2} - \int_{\frac{3}{4}}^1 [6 - 3] - [6y_1 - 3y_1^2] dy_1 \\
&= \frac{1}{2} - \int_{\frac{3}{4}}^1 3 - 6y_1 + 3y_1^2 dy_1 \\
&= \frac{1}{2} - [3y_1 - 3y_1^2 + y_1^3]_{\frac{3}{4}}^1 \\
&= \frac{1}{2} - \left([3 - 3 + 1] - \left[\frac{9}{4} - \frac{27}{16} + \frac{27}{64} \right] \right) \\
&= \frac{1}{2} - \frac{1}{64} = \frac{31}{64}
\end{aligned}$$

Problem 5.16

$$f(y_1, y_2) = \begin{cases} y_1 + y_2 & y_1, y_2 \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$$

(a)

$$\begin{aligned}
P(Y_1 < 1/2, Y_2 > 1/4) &= \int_{\frac{1}{4}}^1 \int_0^{\frac{1}{2}} y_1 + y_2 dy_1 dy_2 \\
&= \int_{\frac{1}{4}}^1 \frac{y_1^2}{2} + y_2 y_1 \Big|_0^{\frac{1}{2}} dy_2 \\
&= \int_{\frac{1}{4}}^1 \frac{1}{8} + \frac{y_2}{2} dy_2 \\
&= \frac{y_2}{8} + \frac{y_2^2}{4} \Big|_{\frac{1}{4}}^1 \\
&= \frac{1}{8} + \frac{1}{4} - \left[\frac{1}{32} + \frac{1}{64} \right] \\
&= \frac{3}{8} - \left[\frac{3}{64} \right] = \frac{21}{64}
\end{aligned}$$

(b)

$$\begin{aligned}P(Y_1 + Y_2 \leq 1) &= P(Y_1 \leq 1 - Y_2) \\&= \int_0^1 \int_0^{1-y_2} y_1 + y_2 dy_1 dy_2 \\&= \int_0^1 \left. \frac{y_1^2}{2} + y_1 y_2 \right|_0^{1-y_2} dy_2 \\&= \int_0^1 \frac{(1-y_2)^2}{2} + (1-y_2)y_2 dy_2 \\&= \int_0^1 \frac{1-2y_2+y_2^2}{2} + y_2 - y_2^2 dy_2 \\&= \int_0^1 \frac{1}{2} - y_2 + \frac{1}{2}y_2^2 + y_2 - y_2^2 dy_2 \\&= \int_0^1 \frac{1}{2} - \frac{1}{2}y_2^2 dy_2 \\&= \frac{1}{2} \int_0^1 1 - y_2^2 dy_2 \\&= \frac{1}{2} \left[y_2 - \frac{y_2^3}{3} \right]_0^1 \\&= \frac{1}{2} \cdot \frac{2}{3} = \frac{1}{3}\end{aligned}$$

Problem 5.24

$$f(y_1, y_2) = \begin{cases} 1 & y_1, y_2 \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$$

(a)

Find the marginal density functions for Y_1, Y_2 .

$$f_1(y_1) = \begin{cases} \int_0^1 1 dy_2 = 1 & 0 \leq y_1 \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$f_2(y_2) = \begin{cases} \int_0^1 1 dy_1 = 1 & 0 \leq y_2 \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

(b)

$$\begin{aligned}P(.3 < Y_1 < .5) &= \int_{.3}^{.5} f_1(y_1) dy_1 \\&= \int_{.3}^{.5} 1 dy_1 \\&= .5 - .3 = .2\end{aligned}$$

$$\begin{aligned}P(.3 < Y_2 < .5) &= \int_{.3}^{.5} f_2(y_2) dy_2 \\&= \int_{.3}^{.5} 1 dy_1 \\&= .5 - .3 = .2\end{aligned}$$

(c)

We know that

$$f(y_1|y_2) = \frac{f(y_1, y_2)}{f_2(y_2)} = \frac{1}{1} = 1$$

So this is defined everywhere for every $y_2 \in [0, 1]$. Otherwise, f_2 is 0 which is clearly undefined.

(d)

From part c, we know that for any $y_2 \in [0, 1]$ that $f(y_1|y_2) = 1$ for all $y_1 \in [0, 1]$.

(e)

$$P(.3 < Y_1 < .5 | Y_2 = .3) = \int_{.3}^{.5} 1 dy_1 = .5 - .3 = .2$$

(f)

$$P(.3 < Y_1 < .5 | Y_2 = .5) = \int_{.3}^{.5} 1 dy_1 = .5 - .3 = .2$$

(g)

The answer does not change at all, this is probably because of the fact that the density function is trivial, and the variables are independent.

Problem 5.38

The variable Y_1 is uniform over $0 \leq y_1 \leq 1$. The variable Y_2 is uniform over $0 \leq y_2 \leq y_1$.

(a)

$$f(y_1, y_2) = \begin{cases} 1 & 0 \leq y_2 \leq y_1 \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

(b)

$$f_1(y_1) = \int_0^{y_1} 1 dy_2 = y_1, \forall y_1 \in [0, 1].$$

$$\begin{aligned} P(Y_2 > 1/4 | Y_1 = 1/2) &= \int_{1/4}^1 f(y_2 | y_1 = 1/2) dy_2 \\ &= \int_{1/4}^1 \frac{1}{(1/2)} dy_2 = 1 - \frac{1}{4} = \frac{3}{4} \end{aligned}$$

Problem 5.48

No, these variables are not independent. The probability $P(Y_1 = 1) = .24$ whereas $P(Y_1 = 1 | Y_2 = 1) = \frac{.02}{.16} = .125$.

Problem 5.68

(a)

We say $g(y_1) = \binom{2}{y_1} .2^{y_1} .8^{2-y_1}$, $h(y_2) = \binom{1}{y_2} .3^{y_2} .7^{1-y_2}$. Then since they are independent the joint density function will be the product of these functions.

$$f(y_1, y_2) = \left(\binom{2}{y_1} .2^{y_1} .8^{2-y_1} \right) \left(\binom{1}{y_2} .3^{y_2} .7^{1-y_2} \right)$$

(b)

$$\begin{aligned}P(Y_1 + Y_2 \leq 1) &= f(0, 0) + f(0, 1) + f(1, 0) \\&= (.8^2)(.7) + (.8^2)(.3) + 2(.2)(.8)(.7) \\&= .864\end{aligned}$$

Problem 5.72

(a)

$$E(Y_1) = np = 2\frac{1}{3} = \frac{2}{3}$$

(b)

$$V(Y_1) = np(1 - p) = 2\frac{2}{3}\frac{1}{3} = \frac{4}{9}$$

(c)

$$E(Y_1 - Y_2) = E(Y_1) - E(Y_2) = \frac{2}{3} - \frac{2}{3} = 0$$

Problem 5.84

(a)

$$\begin{aligned}E(Y_1) &= \frac{p}{q} \\E(Y_2) &= \frac{p}{q} \\E(Y_1 - Y_2) &= E(Y_1) - E(Y_2) = 0\end{aligned}$$

(b)

$$\begin{aligned}E(Y_1^2) &= p\left(\frac{2}{q^2} - \frac{1}{q}\right) \\E(Y_2^2) &= p\left(\frac{2}{q^2} - \frac{1}{q}\right) \\E(Y_1 Y_2) &= E(Y_1)E(Y_2) = \frac{p^2}{q^2}\end{aligned}$$

(c)

$$\begin{aligned}E(Y_1 - Y_2)^2 &= 0 \\V(Y_1 - Y_2) &= E((Y_1 - Y_2)^2) - 0 \\&= E(Y_1^2 - 2Y_1 Y_2 + Y_2^2) \\&= E(Y_1^2) + E(Y_2^2) - 2E(Y_1 Y_2) \\&= 2p\left(\frac{2}{q^2} - \frac{1}{q}\right) - 2\frac{p^2}{q^2} \\&= \frac{2p}{q}\left(\frac{2}{q} - 1 - \frac{p}{q}\right)\end{aligned}$$

Problem 5.94

(a)

$$\begin{aligned} \text{Cov}(U_1, U_2) &= E(U_1 U_2) - E(U_1)E(U_2) \\ &= E([Y_1 + Y_2][Y_1 - Y_2]) - E(Y_1 + Y_2)E(Y_1 - Y_2) \\ &= E(Y_1^2 - Y_2^2) - [E(Y_1) + E(Y_2)][E(Y_1) - E(Y_2)] \\ &= E(Y_1^2) - E(Y_2^2) - E(Y_1)^2 + E(Y_2)^2 \\ &= V(Y_1) - V(Y_2) \end{aligned}$$

(b)

$$\begin{aligned} \rho &= \frac{\text{Cov}(U_1, U_2)}{\sqrt{V(U_1)}\sqrt{V(U_2)}} \\ &= \frac{V(Y_1) - V(Y_2)}{\sqrt{V(Y_1 + Y_2)}\sqrt{V(Y_1 - Y_2)}} \\ &= \frac{V(Y_1) - V(Y_2)}{\sqrt{(V(Y_1) + V(Y_2) + 2\text{Cov}(Y_1, Y_2))(V(Y_1) + V(Y_2) - 2\text{Cov}(Y_1, Y_2))}} \\ &= \frac{V(Y_1) - V(Y_2)}{\sqrt{(V(Y_1) + V(Y_2))(V(Y_1) + V(Y_2))}} \\ &= \frac{V(Y_1) - V(Y_2)}{V(Y_1) + V(Y_2)} \end{aligned}$$

(c)

It is possible that this covariance is equal to 0. This is the case when both variables Y_1 and Y_2 have the same variance.

Problem 5.100

(a)

We have $E(Y_1) = E(Z) = 0$. Then we have $E(Y_2) = E(Z^2) = V(Z) + E(Z)^2 = \sigma^2 + 0^2 = 1$.

(b)

Now we have $E(Y_1 Y_2) = E(Z Z^2) = E(Z^3) = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} x^3 e^{-\frac{x^2}{2}} dx = 0$.

(c)

$$\text{Cov}(Y_1, Y_2) = \text{Cov}(Z, Z^2) = E(Z Z^2) - E(Z)E(Z^2) = 0 - 0(1) = 0$$

(d)

So we know that $P(Y_2 > 1 | Y_1 > 1) = P(Z^2 > 1 | Z > 1) = 1$. However since both, $P(Z^2 > 1)$ and $P(Z > 1)$ are strictly smaller than 1 (in fact most of the normal probability lies between -1 and 1), we know that $P(Z^2 > 1)P(Z > 1)$ must also lie in $(0, 1)$ which does not contain 1, therefore $P(Z^2 > 1 | Z > 1) \neq P(Z^2 > 1)P(Z > 1)$. The two are not independent.