

Probability and Statistics (MA6.101)

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Assignment 5

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Problem 1

We have 12 batteries, of which 3 are new, 4 used and 5 defective. We choose 2 batteries from them without replacement.

X is the number of new batteries chosen and Y the number of used batteries chosen.

Part 1

We want to find $f_{XY}(x, y)$, *i.e.* $P(X = x, Y = y)$. We can do this by noting that the total number of ways to pick 2 batteries from 12 is $\binom{12}{2}$, and the total number of ways to pick x new and y used ones is $\binom{3}{x} \cdot \binom{4}{y}$.

Thus we have

$$f_{XY}(x, y) = \frac{\binom{3}{x} \cdot \binom{4}{y}}{\binom{12}{2}}, x + y \leq 2.$$

Part 2

We want to find $E[X]$, for which we need $f_X(x)$. This can be straightforwardly derived as

$$f_X(x) = \frac{\binom{3}{x}}{\binom{12}{2}},$$

from which we get

$$\begin{aligned} E[X] &= 0 \cdot \left(\frac{1}{\binom{12}{2}} \right) + 1 \cdot \left(\frac{3}{\binom{12}{2}} \right) + 2 \cdot \left(\frac{3}{\binom{12}{2}} \right) \\ &= \frac{9}{66} = \frac{3}{11}. \end{aligned}$$

Problem 2

Part 1

$$\begin{aligned}\Gamma\left(\frac{7}{2}\right) &= \left(\frac{5}{2}\right) \Gamma\left(\frac{5}{2}\right) \\ &= \left(\frac{5}{2}\right) \left(\frac{3}{2}\right) \Gamma\left(\frac{3}{2}\right) \\ &= \left(\frac{5}{2}\right) \left(\frac{3}{2}\right) \left(\frac{1}{2}\right) \Gamma\left(\frac{1}{2}\right) \\ &= \frac{15\sqrt{\pi}}{8}.\end{aligned}$$

Part 2

$$\begin{aligned}I &= \int_0^\infty x^7 e^{-5x} dx \\ &= \int_0^\infty \left(\frac{t}{5}\right)^7 e^{-5\left(\frac{t}{5}\right)} d\left(\frac{t}{5}\right) \\ &= \frac{1}{5^8} \int_0^\infty t^7 e^{-t} dt \\ &= \frac{1}{5^8} \Gamma(8) \\ &= \frac{7!}{5^8}.\end{aligned}$$

Problem 3

We have that Q is a CRV with PDF

$$f_Q(q) = \begin{cases} 6q(1-q) & q \in [0, 1] \\ 0 & \text{otherwise,} \end{cases}$$

and that X is such that

$$P(X = 1 \mid Q = q) = q.$$

We need to find $f_{Q|X}(q \mid x)$ for $x \in \{0, 1\}$ and all q .

First, we will find f_X . Consider $P(X = 1)$; we can find it as follows:

$$\begin{aligned}
P(X = 1) &= \sum_q P(X = 1 \mid Q = q)P(Q = q) \\
&= \int_0^1 P(X = 1 \mid Q = q)f_Q(q)dq \\
&= \int_0^1 6q^2(1 - q)dq \\
&= 6 \left[\frac{q^3}{3} - \frac{q^4}{4} \right]_0^1 \\
&= 6 \cdot \left(\frac{1}{12} \right) = \frac{1}{2}.
\end{aligned}$$

Since X is a Bernoulli RV, we know this means that $P(X = 0) = \frac{1}{2}$. Now, we can calculate

$$\begin{aligned}
f_{Q|X}(q \mid x) &= P(Q = q \mid X = x) \cdot \frac{1}{dq} \\
&= \frac{P(X = x \mid Q = q) \cdot P(Q = q)}{P(X = x)dq}
\end{aligned}$$

For $X = 1$, we have

$$f_{Q|X}(q \mid X = 1) = \frac{q \cdot 6q(1 - q)}{\frac{1}{2}} = 12q^2(1 - q),$$

and for $X = 0$,

$$f_{Q|X}(q \mid X = 0) = \frac{(1 - q) \cdot 6q(1 - q)}{\frac{1}{2}} = 12q(1 - q)^2,$$

since $P(X = x) = \frac{1}{2}$ for all $x \in \{0, 1\}$.

Problem 4

The surface has an infinite number of parallel lines with spacing d , and we throw a needle of length l randomly on it. We need to find the probability that the needle intersects a line.

Since the surface is infinite, we consider the space between two adjacent parallel lines. This is sufficient as the surface merely consists of a number of such spaces.

We will assume that the angle Θ which the needle makes with the vertical and the distance R from the *lower* line (to the centre of the needle) are uniform RVs, *i.e.*,

$$f_{\Theta}(\theta) = \frac{1}{2\pi},$$

$$f_R(r) = \frac{1}{d}.$$

Let the event of the needle intersecting a line be called N . We will first find $f_{N|R}$ and then find $P(N)$ from this.

Suppose that $R = r$. Now there are three cases:

If $r \leq \frac{l}{2}$, then for the needle to intersect the lower line, the maximum value of Θ is $\cos^{-1}\left(\frac{r}{\frac{l}{2}}\right)$. This could be in either direction; therefore we have

$$\begin{aligned} P\left(N \mid R \leq \frac{l}{2}\right) &= P\left(-\cos^{-1}\left(\frac{r}{\frac{l}{2}}\right) \leq \Theta \leq \cos^{-1}\left(\frac{r}{\frac{l}{2}}\right)\right) \\ &= \frac{2\cos^{-1}\left(\frac{2r}{l}\right)}{2\pi} \\ &= \frac{\cos^{-1}\left(\frac{2r}{l}\right)}{\pi}. \end{aligned}$$

If $r \geq d - \frac{l}{2}$, then we obtain an identical result with the distance to the *upper* line, *i.e.*,

$$P\left(N \mid R \geq d - \frac{l}{2}\right) = \frac{\cos^{-1}\left(\frac{2(d-r)}{l}\right)}{\pi}.$$

If $\frac{l}{2} < r < d - \frac{l}{2}$, then it is not possible for the needle to intersect either of the lines, as its centre is more than $\frac{l}{2}$ away from each of them. Thus we have

$$P\left(N \mid \frac{l}{2} < R < d - \frac{l}{2}\right) = 0.$$

Note that the above three cases are mutually exclusive and exhaustive, since we are given that $l < d$.

Thus we have $f_{N|R}$ for all values of R . We can now integrate to find $P(N)$ as follows.

$$\begin{aligned} P(N) &= \sum_r P(N \mid R = r)P(R = r) \\ &= \int_0^d P(N \mid R = r)f_R(r)dr \\ &= \int_0^{\frac{l}{2}} \frac{\cos^{-1}\left(\frac{2r}{l}\right)}{\pi} \frac{dr}{d} \\ &\quad + \int_{d-\frac{l}{2}}^d \frac{\cos^{-1}\left(\frac{2(d-r)}{l}\right)}{\pi} \frac{dr}{d} \\ &\quad + \int_{\frac{l}{2}}^{d-\frac{l}{2}} 0 \frac{dr}{d} \\ &= I_1 + I_2 + 0. \end{aligned}$$

Now, for

$$I_1 = \int_0^{\frac{l}{2}} \frac{\cos^{-1}\left(\frac{2r}{l}\right)}{\pi} \frac{dr}{d},$$

we can take the constants out and substitute $x = \frac{2r}{l}$, which will give us

$$I_1 = \frac{1}{\pi d} \frac{l}{2} \int_0^1 \cos^{-1}(x) dx.$$

Similarly, for

$$I_2 = \int_{d-\frac{l}{2}}^d \frac{\cos^{-1}\left(\frac{2(d-r)}{l}\right)}{\pi} \frac{dr}{d},$$

we can take the constants out and substitute $x = \frac{2(d-r)}{l}$, which gives us

$$\begin{aligned} I_2 &= \frac{1}{\pi d} \int_1^0 \cos^{-1}(x) d\left(-\frac{l}{2}x\right) \\ &= \frac{1}{\pi d} \frac{l}{2} \int_0^1 \cos^{-1}(x) dx. \end{aligned}$$

Thus we have a total of

$$\begin{aligned} P(N) &= \frac{2}{\pi d} \frac{l}{2} \int_0^1 \cos^{-1}(x) dx \\ &= \frac{l}{\pi d} \left[x \cos^{-1}(x) - \sqrt{1-x^2} \right]_0^1 \\ &= \frac{l}{\pi d} [(1 \cdot \cos^{-1}(1) - 0) - (0 \cdot \cos^{-1}(0) - 1)] \\ &= \frac{l}{\pi d} [(0 - 0) - (0 - 1)] \\ &= \frac{l}{\pi d}. \end{aligned}$$

Problem 5

We have two RVs X, Y with range

$$R_{XY} = \{(i, j) \in \mathbb{Z}^2 \mid i, j \geq 0, |i - j| \leq 1\},$$

and joint PMF

$$P_{XY}(i, j) = \frac{1}{6 \cdot 2^{\min(i, j)}}, (i, j) \in R_{XY}.$$

Part 1

The range R_{XY} is represented graphically in the figure.

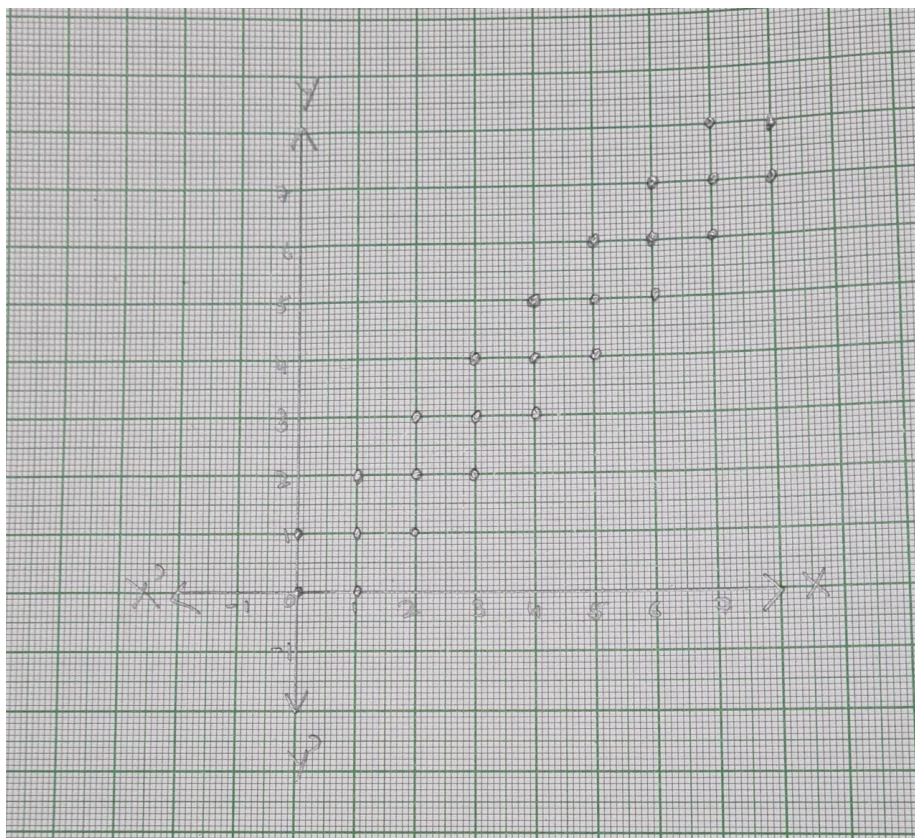


Figure 1: The Range R_{XY}

Part 2

We need to find the marginal PMFs $P_X(i)$ and $P_Y(j)$.

For $P_X(i)$, we must deal with the cases $i = 0$ and $i \geq 1$ separately.

If $i = 0$, then j can only take the values 0 and 1. Summing over these, we get

$$\begin{aligned} P_X(0) &= P_{XY}(0, 0) + P_{XY}(0, 1) \\ &= \frac{1}{6 \cdot 2^0} + \frac{1}{6 \cdot 2^0} \\ &= \frac{1}{3}. \end{aligned}$$

If $i \geq 1$, then $j \in \{i-1, i, i+1\}$. Summing, we get

$$\begin{aligned} P_X(i) &= P_{XY}(i, i-1) + P_{XY}(i, i) + P_{XY}(i, i+1) \\ &= \frac{1}{6 \cdot 2^{i-1}} + \frac{1}{6 \cdot 2^i} + \frac{1}{6 \cdot 2^i} \\ &= \frac{4}{6 \cdot 2^i} \\ &= \frac{1}{3 \cdot 2^{i-1}}. \end{aligned}$$

These can be generalised to a formula for all i as

$$P_X(i) = \frac{1}{3 \cdot 2^{\max(0, i-1)}}.$$

Similarly, for $P_Y(j)$, we deal with the cases $j = 0$ and $j \geq 1$ separately.

If $j = 0$, then i can only take the values 0 and 1. Summing over these, we get

$$\begin{aligned} P_Y(0) &= P_{XY}(0, 0) + P_{XY}(1, 0) \\ &= \frac{1}{6 \cdot 2^0} + \frac{1}{6 \cdot 2^0} \\ &= \frac{1}{3}. \end{aligned}$$

If $j \geq 1$, then $i \in \{j-1, j, j+1\}$. Summing, we get

$$\begin{aligned} P_Y(j) &= P_{XY}(j-1, j) + P_{XY}(j, j) + P_{XY}(j+1, j) \\ &= \frac{1}{6 \cdot 2^{j-1}} + \frac{1}{6 \cdot 2^j} + \frac{1}{6 \cdot 2^j} \\ &= \frac{4}{6 \cdot 2^j} \\ &= \frac{1}{3 \cdot 2^{j-1}}. \end{aligned}$$

Again, these can be generalised to a formula for all j as

$$P_Y(j) = \frac{1}{3 \cdot 2^{\max(0, j-1)}}.$$

We could also have arrived at the second result by the symmetry of the condition.

Part 3

We need to find $P(X = Y \mid X < 2)$.

By Bayes' Law, this is

$$\frac{P(X = Y \cap X < 2)}{P(X < 2)},$$

in which

$$\begin{aligned} P(X = Y \cap X < 2) &= \sum_{i=j, i < 2, (i,j) \in R_{XY}} P_{XY}(i, j) \\ &= P_{XY}(0, 0) + P_{XY}(1, 1) \\ &= \frac{1}{6 \cdot 2^0} + \frac{1}{6 \cdot 2^1} \\ &= \frac{1}{4}, \end{aligned}$$

and

$$\begin{aligned} P(X < 2) &= \sum_{i \in \{0,1\}} P_X(i) \\ &= \frac{1}{3} + \frac{1}{3} \\ &= \frac{2}{3}. \end{aligned}$$

Thus,

$$P(X = Y \mid X < 2) = \frac{\frac{1}{4}}{\frac{2}{3}} = \frac{1}{2}.$$

Part 4

We need to find $P(1 \leq X^2 + Y^2 \leq 5)$.

Proceeding on a case-by-case basis,

$$\begin{aligned} P(1 \leq X^2 + Y^2 \leq 5) &= \sum_{1 \leq i^2 + j^2 \leq 5, (i,j) \in R_{XY}} P_{XY}(i, j) \\ &= \sum_{j \in \{1,2\}} P_{XY}(0, j) + \sum_{j \in \{0,1,2\}} P_{XY}(1, j) + \sum_{j \in \{0,1\}} P_{XY}(2, j) \\ &= \left[2 \left(\frac{1}{6 \cdot 2^0} \right) \right] + \left[\left(\frac{1}{6 \cdot 2^0} \right) + 2 \left(\frac{1}{6 \cdot 2^1} \right) \right] + \left[\left(\frac{1}{6 \cdot 2^0} \right) + \left(\frac{1}{6 \cdot 2^1} \right) \right] \\ &= \left[\frac{1}{3} \right] + \left[\frac{1}{3} \right] + \left[\frac{1}{4} \right] \\ &= \frac{11}{12}. \end{aligned}$$

Part 5

We need to find $P(X = Y)$.

We can sum over R_{XY} with the given condition and find

$$\begin{aligned} P(X = Y) &= \sum_{i=j, (i,j) \in R_{XY}} P_{XY}(i, j) \\ &= \sum_{i \in \mathbb{Z}_0^+} P_{XY}(i, i) \\ &= \sum_{i \in \mathbb{Z}_0^+} \frac{1}{6 \cdot 2^i} \\ &= \frac{1}{6} \sum_{i \in \mathbb{Z}_0^+} \frac{1}{2^i} \\ &= \frac{1}{6}. \end{aligned}$$

Part 6

We have to find $E[X \mid Y = 2]$.

We can do this by first finding $P(X = i \mid Y = 2)$ and then taking the expectation. Note that if $Y = 2$, then $i \in \{1, 2, 3\}$. We then have

$$\begin{aligned} P(X = 1 \mid Y = 2) &= \frac{P_{XY}(1, 2)}{P(Y = 2)} \\ &= 6 \cdot \frac{1}{6 \cdot 2^1} \\ &= \frac{1}{2}; \\ P(X = 2 \mid Y = 2) &= \frac{P_{XY}(2, 2)}{P(Y = 2)} \\ &= 6 \cdot \frac{1}{6 \cdot 2^2} \\ &= \frac{1}{4}; \\ P(X = 3 \mid Y = 2) &= \frac{P_{XY}(3, 2)}{P(Y = 2)} \\ &= 6 \cdot \frac{1}{6 \cdot 2^2} \\ &= \frac{1}{4}. \end{aligned}$$

Thus, we have

$$\begin{aligned} E[X \mid Y = 2] &= 1 \cdot \left(\frac{1}{2}\right) + 2 \cdot \left(\frac{1}{4}\right) + 3 \cdot \left(\frac{1}{4}\right) \\ &= \frac{1}{2} + \frac{1}{2} + \frac{3}{4} \\ &= \frac{7}{4}. \end{aligned}$$

Part 7

We need to find $\text{Var}(X \mid Y = 2)$.

We have found the PMF above. Using it, we obtain

$$\begin{aligned} E[X^2 \mid Y = 2] &= 1^2 \cdot \left(\frac{1}{2}\right) + 2^2 \cdot \left(\frac{1}{4}\right) + 3^2 \cdot \left(\frac{1}{4}\right) \\ &= \frac{1}{2} + 1 + \frac{9}{4} \\ &= \frac{15}{4}, \end{aligned}$$

from which we have

$$\begin{aligned} \text{Var}(X \mid Y = 2) &= E[X^2 \mid Y = 2] - E[X \mid Y = 2]^2 \\ &= \frac{15}{4} - \left(\frac{7}{4}\right)^2 \\ &= \frac{11}{16}. \end{aligned}$$

Problem 6

We know that $X \sim \text{Uniform}(1, 2)$, *i.e.*

$$f_X(x) = 1,$$

and given $X = x$, $Y \sim \text{Exponential}(x)$, *i.e.*,

$$f_{Y|X}(Y \mid X = x) = xe^{-xy}.$$

First, we find the PDF of Y . We can proceed as follows.

$$\begin{aligned}
f_Y(y) &= \sum f_{Y|X}(Y | X = x)P(X = x) \\
&= \int_1^2 f_{Y|X}(Y | X = x)f_X(x)dx \\
&= \int_1^2 xe^{-xy} \cdot 1dx \\
&= \left[-\frac{x}{y}e^{-xy} - \frac{1}{y^2}e^{-xy} \right]_1^2 \\
&= \left[e^{-xy} \left(x + \frac{1}{y} \right) \right]_1^2 \\
&= e^{-y} \left(1 + \frac{1}{y} \right) - e^{-2y} \left(2 + \frac{1}{y} \right).
\end{aligned}$$

Part 1

We want to find $E[Y]$.

We use the PDF found above and integrate as follows.

$$\begin{aligned}
E[Y] &= \int_0^\infty yf_Y(y)dy \\
&= \int_0^\infty \left[ye^{-y} \left(1 + \frac{1}{y} \right) - ye^{-2y} \left(2 + \frac{1}{y} \right) \right] dy \\
&= I_1 + I_2 - I_3 - I_4.
\end{aligned}$$

Then,

$$\begin{aligned}
I_1 &= \int_0^\infty ye^{-y}dy \\
&= [-ye^{-y} - e^{-y}]_0^\infty \\
&= 0 - (-1) = 1.
\end{aligned}$$

$$\begin{aligned}
I_2 &= \int_0^\infty e^{-y}dy \\
&= [-e^{-y}]_0^\infty \\
&= 0 - (-1) = 1.
\end{aligned}$$

$$\begin{aligned}
I_3 &= \int_0^{\infty} 2ye^{-2y} dy \\
&= \frac{1}{2} [-(2y)e^{-2y} - e^{-2y}]_0^{\infty} \\
&= \frac{1}{2} (0 - (-1)) = \frac{1}{2}.
\end{aligned}$$

$$\begin{aligned}
I_4 &= \int_0^{\infty} e^{-2y} dy \\
&= \frac{1}{2} [-e^{-2y}]_0^{\infty} \\
&= \frac{1}{2} (0 - (-1)) = \frac{1}{2}.
\end{aligned}$$

Therefore,

$$E[Y] = 1 + 1 - \frac{1}{2} - \frac{1}{2} = 1.$$

Part 2

We want to find $\text{Var}(Y)$.

We will use the identity $\text{Var}(Y) = E[Y^2] - E[Y]^2$. To find $E[Y^2]$,

$$\begin{aligned}
E[Y^2] &= \int_0^{\infty} y^2 f_Y(y) dy \\
&= \int_0^{\infty} \left[y^2 e^{-y} \left(1 + \frac{1}{y} \right) - y^2 e^{-2y} \left(2 + \frac{1}{y} \right) \right] dy \\
&= I_1 + I_2 - I_3 - I_4.
\end{aligned}$$

Then,

$$\begin{aligned}
I_1 &= \int_0^{\infty} y^2 e^{-y} dy \\
&= \left[-y^2 e^{-y} + \int 2ye^{-y} \right]_0^{\infty} \\
&= \left[-y^2 e^{-y} - \int 2ye^{-y} - 2e^{-y} \right]_0^{\infty} \\
&= 0 - (-2) = 2.
\end{aligned}$$

$$I_2 = \int_0^{\infty} ye^{-y} dy = 1.$$

$$\begin{aligned}
I_3 &= \int_0^\infty 2y^2 e^{-2y} dy \\
&= \frac{1}{4} \left[-(2y)^2 e^{-2y} - \int 2ye^{-2y} - 2e^{-2y} \right]_0^\infty \\
&= \frac{1}{4} (0 - (-2)) = \frac{1}{2}.
\end{aligned}$$

$$\begin{aligned}
I_4 &= \int_0^\infty ye^{-2y} dy \\
&= \frac{1}{4} \int_0^\infty (2y)e^{-2y} d(2y) \\
&= \frac{1}{4} [-e^{-2y}]_0^\infty \\
&= \frac{1}{4} (0 - (-1)) = \frac{1}{4}.
\end{aligned}$$

Therefore,

$$E[Y^2] = 2 + 1 - \frac{1}{4} - \frac{1}{2} = \frac{9}{4}.$$

Thus, we have

$$\begin{aligned}
\text{Var}(Y) &= E[Y^2] - E[Y]^2 \\
&= \frac{9}{4} - 1^2 \\
&= \frac{5}{4}.
\end{aligned}$$

Problem 7

We have that $X \sim N(\mu_X, \sigma_X^2)$ and $Y \sim N(\mu_Y, \sigma_Y^2)$ are independent variables. We need to show that

$$X + Y \sim N(\mu_X + \mu_Y, \sigma_X^2 + \sigma_Y^2).$$

Let $Z = X + Y$. We wish to find $f_Z(z)$, which we can do as follows.

$$\begin{aligned}
f_Z(z) &= P(X + Y = z) \\
&= \sum_{x+y=z} P(X = x)P(Y = y) \\
&= \sum_x P(X = x)P(Y = z - x) \\
&= \int_{-\infty}^{\infty} f_X(x)f_Y(z - x)dx \\
&= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma_Y} e^{\left[-\frac{(z-x-\mu_Y)^2}{2\sigma_Y^2}\right]} \frac{1}{\sqrt{2\pi}\sigma_X} e^{\left[-\frac{(x-\mu_X)^2}{2\sigma_X^2}\right]} dx \\
&= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma_X\sqrt{2\pi}\sigma_Y} \exp\left[-\frac{\sigma_X^2(z-x-\mu_Y)^2 + \sigma_Y^2(x-\mu_X)^2}{2\sigma_X^2\sigma_Y^2}\right] dx \\
&= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma_Z\sqrt{2\pi}\left(\frac{\sigma_X\sigma_Y}{\sigma_Z}\right)} \cdot \exp(E),
\end{aligned}$$

where

$$\sigma_Z = \sqrt{\sigma_X^2 + \sigma_Y^2},$$

and

$$\begin{aligned}
E &= -\frac{\sigma_X^2(z-x-\mu_Y)^2 + \sigma_Y^2(x-\mu_X)^2}{2\sigma_X^2\sigma_Y^2} \\
&= -\frac{x^2(\sigma_X^2 + \sigma_Y^2) - 2x(\sigma_Y^2\mu_X + \sigma_X^2(z-\mu_Y)) + \sigma_X^2(z-\mu_Y)^2 + \sigma_Y^2\mu_X^2}{2\sigma_X^2\sigma_Y^2} \\
&= -\frac{x - 2x\frac{\sigma_Y^2\mu_X + \sigma_X^2(z-\mu_Y)}{\sigma_Z^2} + \frac{\sigma_X^2(z-\mu_Y)^2 + \sigma_Y^2\mu_X^2}{\sigma_Z^2}}{2\left(\frac{\sigma_X\sigma_Y}{\sigma_Z}\right)^2} \\
&= -\frac{\left(x - \frac{\sigma_Y^2\mu_X + \sigma_X^2(z-\mu_Y)}{\sigma_Z^2}\right)^2 - \left(\frac{\sigma_Y^2\mu_X + \sigma_X^2(z-\mu_Y)}{\sigma_Z^2}\right)^2 + \frac{\sigma_X^2(z-\mu_Y)^2 + \sigma_Y^2\mu_X^2}{\sigma_Z^2}}{2\left(\frac{\sigma_X\sigma_Y}{\sigma_Z}\right)^2} \\
&= -\frac{\left(x - \frac{\sigma_Y^2\mu_X + \sigma_X^2(z-\mu_Y)}{\sigma_Z^2}\right)^2}{2\left(\frac{\sigma_X\sigma_Y}{\sigma_Z}\right)^2} - \frac{\sigma_Z^2(\sigma_Y^2\mu_X^2 + \sigma_X^2(z-\mu_Y)^2) - (\sigma_Y^2\mu_X + \sigma_X^2(z-\mu_Y))^2}{2\sigma_Z^2(\sigma_X\sigma_Y)^2} \\
&= E_1 + E_2.
\end{aligned}$$

We will leave E_1 as it is, and simplify E_2 .

$$\begin{aligned}
E_2 &= -\frac{\sigma_Z^2(\sigma_Y^2\mu_X^2 + \sigma_X^2(z - \mu_Y)^2) - (\sigma_Y^2\mu_X + \sigma_X^2(z - \mu_Y))^2}{2\sigma_Z^2(\sigma_X\sigma_Y)^2} \\
&= -\frac{(\sigma_X^2 + \sigma_Y^2)(\sigma_Y^2\mu_X^2 + \sigma_X^2(z - \mu_Y)^2) - (\sigma_Y^2\mu_X + \sigma_X^2(z - \mu_Y))^2}{2\sigma_Z^2(\sigma_X\sigma_Y)^2} \\
&= -\frac{(\sigma_X^2 + \sigma_Y^2)(\sigma_Y^2\mu_X^2 + \sigma_X^2(z - \mu_Y)^2) - \sigma_Y^4\mu_X^2 - \sigma_X^4(z - \mu_Y)^2 - 2\sigma_X^2\sigma_Y^2\mu_X(z - \mu_Y)}{2\sigma_Z^2(\sigma_X\sigma_Y)^2} \\
&= -\frac{\sigma_X^2\sigma_Y^2(\mu_X^2 + (z - \mu_Y)^2 - 2\mu_X(z - \mu_Y)) + \sigma_X^4((z - \mu_Y)^2 - (z - \mu_Y)^2) + \sigma_Y^4(\mu_X^2 - \mu_X^2)}{2\sigma_Z^2(\sigma_X\sigma_Y)^2} \\
&= -\frac{((z - \mu_Y) - \mu_X)^2}{2\sigma_Z^2} \\
&= -\frac{(z - (\mu_X + \mu_Y))^2}{2\sigma_Z^2}.
\end{aligned}$$

Thus our initial equation becomes

$$\begin{aligned}
f_Z(z) &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma_Z\sqrt{2\pi}\left(\frac{\sigma_X\sigma_Y}{\sigma_Z}\right)} \cdot \exp(E) \\
&= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma_Z\sqrt{2\pi}\left(\frac{\sigma_X\sigma_Y}{\sigma_Z}\right)} \cdot \exp(E_1 + E_2) \\
&= \frac{1}{\sqrt{2\pi}\sigma_Z} \cdot \exp(E_2) \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\left(\frac{\sigma_X\sigma_Y}{\sigma_Z}\right)} \exp(E_1) \\
&= \frac{1}{\sqrt{2\pi}\sigma_Z} \cdot \exp\left[-\frac{(z - (\mu_X + \mu_Y))^2}{2\sigma_Z^2}\right] \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\left(\frac{\sigma_X\sigma_Y}{\sigma_Z}\right)} \exp\left[-\frac{\left(x - \frac{\sigma_Y^2\mu_X + \sigma_X^2(z - \mu_Y)}{\sigma_Z^2}\right)^2}{2\left(\frac{\sigma_X\sigma_Y}{\sigma_Z}\right)^2}\right] dx \\
&= \frac{1}{\sqrt{2\pi}\sigma_Z} \cdot \exp\left[-\frac{(z - (\mu_X + \mu_Y))^2}{2\sigma_Z^2}\right],
\end{aligned}$$

since the integral is the normalised Gaussian integral, which is known to evaluate to 1. We can be sure of this as it is the integral of the PDF of a variable with distribution

$$N\left(\frac{\sigma_Y^2\mu_X + \sigma_X^2(z - \mu_Y)}{\sigma_Z^2}, \left(\frac{\sigma_X\sigma_Y}{\sigma_Z}\right)^2\right).$$

Now from the form of $f_Z(z)$, we can conclude that $Z \sim N(\mu_X + \mu_Y, \sigma_X^2 + \sigma_Y^2)$, QED.

Problem 8

We are given that $X_1 \sim N(2, 3)$ and $X_2 \sim N(1, 4)$ are two independent normal RVs.

Part 1

We need the distribution of $Y = 2X_1 + 3X_2$.

We know that if Y is a normal variable and $X = \sigma Y + \mu$, then X is a normal variable such that $E[X] = \sigma E[Y] + \mu$, and $\text{Var}(X) = \sigma^2 \text{Var}(Y)$.

Using this property, we can conclude that $2X_1 \sim N(4, 12)$, and $3X_2 \sim N(3, 36)$.

From the result of Problem 7 above, we then know that $Y \sim N(7, 48)$.

Part 2

We need the distribution of $Y = X_1 - X_2$.

Suppose we have an RV $X = -X_2$. Then its PDF is

$$\begin{aligned} f_X(x) &= f_{X_2}(-x) \\ &= \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(-x-\mu)^2}{2\sigma^2}} \\ &= \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-(-\mu))^2}{2\sigma^2}}, \end{aligned}$$

which means that $X \sim N(-\mu, \sigma)$.

Using this property, we can combine X_1 and $-X_2$ as we did in Part 1, getting $Y \sim N(1, 7)$.

Problem 9

We have defined

$$D = \{(x, y) \mid x^2 + y^2 \leq 1\},$$

and the joint PDF of two RVs X, Y as

$$f_{XY}(x, y) = \begin{cases} c & (x, y) \in D \\ 0 & \text{otherwise,} \end{cases}$$

for some constant c .

Part 1

We need the value of the constant c .

As the probability is constant all over the disc, to find the total probability, we only need to multiply c with the area $\pi(1)^2 = \pi$ of the disc. This should be equal to 1; thus we get $c = \frac{1}{\pi}$.

Part 2

We want to obtain the marginal PDFs $f_X(x)$ and $f_Y(y)$.

First, we will find $f_X(x)$. By definition, we have

$$\begin{aligned} f_X(x) &= \sum_{x^2+y^2 \leq 1} f_{XY}(x, y) \\ &= \int_{x^2+y^2 \leq 1} \frac{1}{\pi} dy \\ &= \int_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} \frac{dy}{\pi} \\ &= \frac{2\sqrt{1-x^2}}{\pi}. \end{aligned}$$

We can proceed in an exactly similar way to obtain $f_Y(y)$:

$$\begin{aligned} f_Y(y) &= \sum_{x^2+y^2 \leq 1} f_{XY}(x, y) \\ &= \int_{x^2+y^2 \leq 1} \frac{1}{\pi} dx \\ &= \int_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}} \frac{dx}{\pi} \\ &= \frac{2\sqrt{1-y^2}}{\pi}. \end{aligned}$$

We could also have used symmetry to obtain the same result.

Part 3

We want to find the conditional PDF of X given $Y = y$, for any $y \in [-1, 1]$.

Using Bayes' Law, we know that

$$\begin{aligned} P(X = x \mid Y = y) &= \frac{P(X = x) \cap P(Y = y)}{P(Y = y)} \\ &= \frac{f_{XY}(x, y) dx dy}{f_Y(y) dy} \\ \Rightarrow f_{X|Y}(X = x \mid Y = y) dx &= \frac{f_{XY}(x, y) dx dy}{P(Y = y) dy} \\ \Rightarrow f_{X|Y}(X = x \mid Y = y) &= \frac{f_{XY}(x, y)}{f_Y(y)}. \end{aligned}$$

This gives us

$$\begin{aligned} f_{X|Y}(X = x \mid Y = y) &= \frac{\frac{1}{\pi}}{\frac{2\sqrt{1-y^2}}{\pi}} \\ &= \frac{1}{2\sqrt{1-y^2}}. \end{aligned}$$

Part 4

We want to know if X, Y are independent.

We know that they are independent if $f_{XY}(x, y) = f_X(x)f_Y(y)$ for all $(x, y) \in D$. The LHS of this, as we know, is $\frac{1}{\pi}$, and the RHS is

$$\frac{2\sqrt{1-x^2}}{\pi} \frac{2\sqrt{1-y^2}}{\pi},$$

which is clearly not constant. Thus the variables are not independent.

Problem 10

We are given two RVs X, Y with the joint PDF

$$f_{XY}(x, y) = \begin{cases} xe^{-x(1+y)} & x, y \geq 0 \\ 0 & \text{otherwise,} \end{cases}$$

which represent the lifetimes in years of two components of a laptop.

Part 1

We need to find the marginal PDFs $f_X(x)$ and $f_Y(y)$.

Proceeding as before, for $f_X(x)$,

$$\begin{aligned} f_X(x) &= \sum_y f_{XY}(x, y) \\ &= \int_0^\infty xe^{-x(1+y)} dy \\ &= xe^{-x} \int_0^\infty e^{-xy} dy \\ &= \frac{1}{x} xe^{-x} \int_0^\infty e^{-t} dt \\ &= e^{-x} [-e^{-t}]_0^\infty \\ &= e^{-x}. \end{aligned}$$

Similarly, for $f_Y(y)$,

$$\begin{aligned}
f_Y(y) &= \sum_x f_{XY}(x, y) \\
&= \int_0^\infty x e^{-x(1+y)} dx \\
&= \left[\frac{x}{1+y} (-e^{-x(1+y)}) + \frac{1}{1+y} \int e^{-x(1+y)} dx \right]_0^\infty \\
&= \left[-e^{-x(1+y)} \left(\frac{x}{1+y} + \frac{1}{(1+y)^2} \right) \right]_0^\infty \\
&= 0 - \left(-\frac{1}{(1+y)^2} \right) \\
&= \frac{1}{(1+y)^2}.
\end{aligned}$$

Part 2

We need to find the probability that the lifetime of at least one component exceeds one year, *i.e.*, $P(X > 1 \cup Y > 1)$.

We know from the Inclusion-Exclusion Principle that

$$P(X > 1 \cup Y > 1) = P(X > 1) + P(Y > 1) - P(X > 1 \cap Y > 1).$$

Now, we can find $P(X > 1)$ from $f_X(x)$ by integrating:

$$\begin{aligned}
P(X > 1) &= \int_1^\infty f_X(x) dx \\
&= \int_1^\infty e^{-x} dx \\
&= [-e^{-x}]_1^\infty \\
&= 0 - (-e^{-1}) = \frac{1}{e},
\end{aligned}$$

and similarly for $P(Y > 1)$:

$$\begin{aligned}
P(Y > 1) &= \int_1^\infty f_Y(y) dy \\
&= \int_1^\infty \frac{dy}{(1+y)^2} \\
&= \int_2^\infty \frac{dt}{t^2} \\
&= \left[-\frac{1}{t} \right]_2^\infty \\
&= 0 - \left(-\frac{1}{2} \right) = \frac{1}{2},
\end{aligned}$$

and finally for the intersection:

$$\begin{aligned}
P(X > 1 \cap Y > 1) &= \int_1^\infty \int_1^\infty x e^{-x(1+y)} dy dx \\
&= \int_1^\infty x e^{-x} \int_1^\infty e^{-xy} dy dx \\
&= \int_1^\infty e^{-x} [-e^{-xy}]_{y=1}^\infty dx \\
&= \int_1^\infty 2e^{-x} dx \\
&= -2[e^{-x}]_1^\infty \\
&= -2(0 - e^{-1}) = \frac{2}{e}.
\end{aligned}$$

Thus, overall we have

$$\begin{aligned}
P(X > 1 \cup Y > 1) &= \frac{1}{e} + \frac{1}{2} - \frac{2}{e} \\
&= \frac{1}{2} - \frac{1}{e} \\
&\approx 0.132
\end{aligned}$$