

3 Signal & Systems (Repetition)

3.5

The hint and answers in the problem collection provides the full solution to this problem.

1 Stochastic Variables (Repetition)

1.1

a. Check that the product of probabilities are the probabilities of the intersections:

$$\begin{aligned}\Pr\{A\} \cdot \Pr\{B\} &= \frac{1}{2} \cdot \frac{1}{3} = \frac{1}{6} = \Pr\{A \cap B\} \\ \Pr\{A\} \cdot \Pr\{C\} &= \frac{1}{2} \cdot \frac{1}{4} = \frac{1}{8} = \Pr\{A \cap C\} \\ \Pr\{B\} \cdot \Pr\{C\} &= \frac{1}{3} \cdot \frac{1}{4} = \frac{1}{12} = \Pr\{B \cap C\} \\ \Pr\{A\} \cdot \Pr\{B\} \cdot \Pr\{C\} &= \frac{1}{2} \cdot \frac{1}{3} \cdot \frac{1}{4} = \frac{1}{24} \neq \Pr\{A \cap B \cap C\}\end{aligned}$$

Thus, the events A , B och C are dependent, but they are pairwise independent.

b.

$$\Pr\{A \cup B\} = \Pr\{A\} + \Pr\{B\} - \Pr\{A \cap B\} = \frac{1}{2} + \frac{1}{3} - \frac{1}{6} = \frac{2}{3}$$

c.

$$\begin{aligned}\Pr\{A \cup B \cup C\} &= \Pr\{A\} + \Pr\{B\} + \Pr\{C\} \\ &\quad - \Pr\{A \cap B\} - \Pr\{A \cap C\} - \Pr\{B \cap C\} \\ &\quad + \Pr\{A \cap B \cap C\} \\ &= \frac{1}{2} + \frac{1}{3} + \frac{1}{4} - \frac{1}{6} - \frac{1}{8} - \frac{1}{12} + \frac{1}{24} = \frac{37}{48}\end{aligned}$$

d.

$$\begin{aligned}\Pr\{B \mid C\} &= \frac{\Pr\{B \cap C\}}{\Pr\{C\}} = \frac{1/12}{1/4} = \frac{1}{3} \\ \Pr\{A \mid B \cap C\} &= \frac{\Pr\{A \cap B \cap C\}}{\Pr\{B \cap C\}} = \frac{1/24}{1/12} = \frac{1}{2}\end{aligned}$$

Answer:

a. The events A , B och C are dependent, but they are pairwise independent.

b. $\Pr\{A \cup B\} = \frac{2}{3}$ c. $\Pr\{A \cup B \cup C\} = \frac{37}{48}$ d. $\Pr\{B \mid C\} = \frac{1}{3}$, $\Pr\{A \mid B \cap C\} = \frac{1}{2}$

1.2

Let $M = (M_1 M_2 M_3)$ be the input. Then according to the problem description, we have

$$\begin{aligned}\Pr\{M = (000)\} &= \Pr\{M = (110)\} = \Pr\{M = (101)\} = \Pr\{M = (011)\} = \frac{1}{4} \\ \Pr\{M = (111)\} &= \Pr\{M = (001)\} = \Pr\{M = (010)\} = \Pr\{M = (100)\} = 0\end{aligned}$$

Let $R = (R_1 R_2 R_3)$ be the output. Finally, let B_i be the event $R_i = 0$.

a.

$$\begin{aligned}\Pr\{B_1\} &= \Pr\{M = (000)\} \cdot (1 - p) + \Pr\{M = (011)\} \cdot (1 - p) \\ &\quad + \Pr\{M = (110)\} \cdot p + \Pr\{M = (101)\} \cdot p \\ &= 2 \cdot \frac{1}{4}(1 - p) + 2 \cdot \frac{1}{4}p = \frac{1}{2}\end{aligned}$$

Note: Obviously, we have $\Pr\{B_1\} = \Pr\{B_2\} = \Pr\{B_3\}$.

b.

$$\begin{aligned}\Pr\{R = (110)\} &= \Pr\{M = (000)\} \cdot p^2(1 - p) + \Pr\{M = (011)\} \cdot p^2(1 - p) \\ &\quad + \Pr\{M = (110)\} \cdot (1 - p)^3 + \Pr\{M = (101)\} \cdot p^2(1 - p) \\ &= \frac{3}{4}p^2(1 - p) + \frac{1}{4}(1 - p)^3 \approx 0.189\end{aligned}$$

c. Using Bayes's rule, we get

$$\begin{aligned}\Pr\{M = (110) \mid R = (110)\} &= \frac{\Pr\{M = (110)\} \Pr\{R = (110) \mid M = (110)\}}{\Pr\{R = (110)\}} \\ &= \frac{\frac{1}{4}(1 - p)^3}{0.189} \approx 0.9643\end{aligned}$$

0.964

d. Check that the product of probabilities are the probabilities of the intersections:

$$\begin{aligned}\Pr\{B_1 \cap B_2\} &= \Pr\{M = (000)\} \cdot (1 - p)^2 + \Pr\{M = (011)\} \cdot p(1 - p) \\ &\quad + \Pr\{M = (110)\} \cdot p^2 + \Pr\{M = (101)\} \cdot p(1 - p) = \frac{1}{4} \\ \Pr\{B_1\} \Pr\{B_2\} &= \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}\end{aligned}$$

Similarly for $\Pr\{B_1 \cap B_3\}$ and $\Pr\{B_2 \cap B_3\}$. For all three events, we have

$$\begin{aligned}\Pr\{B_1 \cap B_2 \cap B_3\} &= \Pr\{M = (000)\} \cdot (1 - p)^3 + \Pr\{M = (011)\} \cdot p^2(1 - p) \\ &\quad + \Pr\{M = (110)\} \cdot p^2(1 - p) + \Pr\{M = (101)\} \cdot p^2(1 - p) \neq \frac{1}{8}\end{aligned}$$

$$\Pr\{B_1\} \Pr\{B_2\} \Pr\{B_3\} = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{8}$$

Thus, the events B_1, B_2, B_3 are dependent, but they are pairwise independent.

Answer:

a. $\Pr\{B_1\} = \frac{1}{2}$ **b.** $\Pr\{R = (110)\} = 0.189$ **c.** 0.964

d. B_1, B_2, B_3 are dependent, but they are pairwise independent.

1.3

$$\Pr\{X < 0.99\} = 1 - \Pr\{X \geq 0.99\} = 1 - Q\left(\frac{0.99 - m_X}{\sigma_X}\right) = 1 - Q(199),$$

$$\Pr\{X < -0.99\} = 1 - \Pr\{X \geq -0.99\} = 1 - Q\left(\frac{-0.99 - m_X}{\sigma_X}\right) = 1 - Q(1),$$

$$\Pr\{X > -0.99\} = Q\left(\frac{-0.99 - m_X}{\sigma_X}\right) = Q(1),$$

$$\begin{aligned}\Pr\{-1.3 < X < -1.03\} &= Q\left(\frac{-1.3 - m_X}{\sigma_X}\right) - Q\left(\frac{-1.03 - m_X}{\sigma_X}\right) = Q(-30) - Q(-3) \\ &= (1 - Q(30)) - (1 - Q(3)) = Q(3) - Q(30).\end{aligned}$$

Answer:

$$\Pr\{X < 0.99\} = 1 - Q(199)$$

$$\Pr\{X < -0.99\} = 1 - Q(1)$$

$$\Pr\{X > -0.99\} = Q(1)$$

$$\Pr\{-1.3 < X < -1.03\} = Q(3) - Q(30)$$

1.4

Sent signal: $X \in \{\pm 1\}$, equally probable.

Added noise: N , uniformly distributed on $[-2, 2)$.

Received signal: $Y = X + N$

We get conditional PDFs:

$$f_{Y|X}(y | -1) = \begin{cases} 1/4, & -3 \leq y < 1, \\ 0, & \text{elsewhere} \end{cases}$$

$$f_{Y|X}(y | 1) = \begin{cases} 1/4, & -1 \leq y < 3, \\ 0, & \text{elsewhere} \end{cases}$$

Error probability: $P_e = \Pr\{\hat{X} \neq X\}$, where \hat{X} is our estimate according to the chosen decision rule.

General expression:

$$\begin{aligned} P_e &= \Pr\{\hat{X} \neq X\} = \Pr\{X = -1\}\Pr\{\hat{X} = 1 | X = -1\} + \Pr\{X = 1\}\Pr\{\hat{X} = -1 | X = 1\} \\ &= \frac{1}{2}(\Pr\{\hat{X} = 1 | X = -1\} + \Pr\{\hat{X} = -1 | X = 1\}) \end{aligned}$$

a. Given rule: $\hat{X} = \begin{cases} -1, & Y \leq 0, \\ 1, & Y > 0. \end{cases}$

$$\Pr\{\hat{X} = 1 | X = -1\} = \Pr\{\hat{Y} > 0 | X = -1\} = \int_0^1 \frac{1}{4} dy = \frac{1}{4},$$

$$\Pr\{\hat{X} = -1 | X = 1\} = \Pr\{\hat{Y} \leq 0 | X = 1\} = \int_{-1}^0 \frac{1}{4} dy = \frac{1}{4}.$$

Plug that into the general expression, and we get

$$P_e = \frac{1}{2} \left(\frac{1}{4} + \frac{1}{4} \right) = \frac{1}{4}$$

b. Given rule: $\hat{X} = \begin{cases} -1, & Y \leq -1, \\ 1, & Y > -1. \end{cases}$

$$\Pr\{\hat{X} = 1 | X = -1\} = \Pr\{\hat{Y} > -1 | X = -1\} = \int_{-1}^1 \frac{1}{4} dy = \frac{1}{2},$$

$$\Pr\{\hat{X} = -1 | X = 1\} = \Pr\{\hat{Y} \leq -1 | X = 1\} = 0.$$

Plug that into the general expression, and we get

$$P_e = \frac{1}{2} \left(\frac{1}{2} + 0 \right) = \frac{1}{4}$$

c. Given rule: $\hat{X} = \begin{cases} -1, & Y \leq -1, \\ 1, & Y > 1, \\ \pm 1, & \text{elsewhere.} \end{cases}$, where ± 1 is chosen by the toss of a fair coin.

$$\begin{aligned} \Pr\{\hat{X}=1 \mid X=-1\} &= \Pr\{\hat{Y} > 1 \mid X=-1\} + \frac{1}{2}\Pr\{-1 < Y \leq 1 \mid X=-1\} \\ &= 0 + \frac{1}{2} \int_{-1}^1 \frac{1}{4} dy = \frac{1}{4}, \end{aligned}$$

$$\begin{aligned} \Pr\{\hat{X}=-1 \mid X=1\} &= \Pr\{\hat{Y} \leq -1 \mid X=1\} + \frac{1}{2}\Pr\{-1 < Y \leq 1 \mid X=1\} \\ &= 0 + \frac{1}{2} \int_{-1}^1 \frac{1}{4} dy = \frac{1}{4}. \end{aligned}$$

Plug that into the general expression, and we get

$$P_e = \frac{1}{2} \left(\frac{1}{4} + \frac{1}{4} \right) = \frac{1}{4}$$

Answer:

1.5

a. We have

$$1 = \int_{-\infty}^{\infty} f_X(x) dx = 2 \int_0^{\infty} e^{-cx} dx = 2 \left[\frac{e^{-cx}}{-c} \right]_0^{\infty} = \frac{2}{c},$$

from which we get $c = 2$.

b. First we observe that all odd moments ($E\{X^n\}$ for n odd) are zero since $f_X(x)$ is even. For n positive and even, we have

$$E\{X^n\} = \int_{-\infty}^{\infty} x^n f_X(x) dx = \int_{-\infty}^{\infty} x^n e^{-2|x|} dx = 2 \int_0^{\infty} x^n e^{-2x} dx.$$

We attack this expression using partial integration, where x^n is one of the functions and e^{-2x} is the other. We get

$$E\{X^n\} = 2 \left[x^n \cdot \frac{e^{-2x}}{-2} \right]_0^{\infty} - 2 \int_0^{\infty} nx^{n-1} \cdot \frac{e^{-2x}}{-2} dx = \frac{n}{2} \cdot 2 \int_0^{\infty} x^{n-1} e^{-2x} dx.$$

another partial integration, and we get

$$\begin{aligned} E\{X^n\} &= n \left[x^{n-1} \cdot \frac{e^{-2x}}{-2} \right]_0^{\infty} - \frac{n}{2} \cdot 2 \int_0^{\infty} (n-1)x^{n-2} \cdot \frac{e^{-2x}}{-2} dx \\ &= \frac{n}{2} \cdot \frac{n-1}{2} \cdot 2 \int_0^{\infty} x^{n-2} e^{-2x} dx = \frac{n}{2} \cdot \frac{n-1}{2} \cdot E\{X^{n-2}\}. \end{aligned}$$

Doing this $n/2$ times, we end up with

$$E\{X^n\} = \frac{n}{2} \cdot \frac{n-1}{2} \cdots \frac{2}{2} \cdot \frac{1}{2} \cdot E\{X^0\}.$$

Obviously, we have

$$E\{X^0\} = E\{1\} = 1.$$

Totally, this gives us

$$E\{X^n\} = \begin{cases} 0, & n \text{ odd}, \\ \frac{n!}{2^n}, & n \text{ even}. \end{cases}$$

c. We want to determine $f_Z(z)$ for $Z = X + Y$, where X and Y are independent with $f_X(x) = f_Y(x) = e^{-2|x|}$. That is a line integral of $f_{X,Y}(x, y)$ along the line $x + y = z$, i.e.

$$f_Z(z) = \int_{-\infty}^{\infty} f_{X,Y}(x, z-x) dx$$

X and Y are independent, which gives us

$$f_Z(z) = \int_{-\infty}^{\infty} f_X(x)f_Y(z-x) dx = (f_X * f_Y)(z)$$

To calculate this convolution, we need to consider the two cases $z > 0$ and $z \leq 0$. For $z > 0$, we have

$$\begin{aligned} f_Z(z) &= \int_{-\infty}^0 e^{2x} e^{-2(z-x)} dx + \int_0^z e^{-2x} e^{-2(z-x)} dx + \int_z^{\infty} e^{-2x} e^{2(z-x)} dx \\ &= e^{-2z} \int_{-\infty}^0 e^{4x} dx + e^{-2z} \int_0^z dx + e^{2z} \int_z^{\infty} e^{-4x} dx \\ &= e^{-2z} \frac{1}{4} + e^{-2z} \cdot z + e^{2z} \frac{1}{4} = e^{-2z} \left(\frac{1}{2} + z \right) \end{aligned}$$

Similarly for $z \leq 0$, we have

$$f_Z(z) = e^{2z} \left(\frac{1}{2} - z \right).$$

Totally, we can write this as

$$f_Z(z) = e^{-2|z|} \left(\frac{1}{2} + |z| \right)$$

for all z .

d. To determine $E\{Z^2\}$, we use the results from above including the independence:

$$\begin{aligned} E\{Z^2\} &= E\{(X+Y)^2\} = E\{X^2\} + 2E\{XY\} + E\{Y^2\} \\ &= 2E\{X^2\} + 2E\{XY\} = 2\frac{2!}{2^2} + 2 \cdot 0 = 1. \end{aligned}$$

Answer:

a. $c=2$

$$\text{b. } E\{X^n\} = \begin{cases} 0, & n \text{ odd} \\ \frac{n!}{2^n}, & n \text{ even} \end{cases}$$

$$\text{c. } f_Z(z) = e^{-2|z|} \left(\frac{1}{2} + |z| \right)$$

d. $E\{Z^2\} = 1$

1.6

- a. The area of a circle of radius 1 is π . The integral of the PDF is always 1. Thus, we have $B = \frac{1}{\pi}$.
- b. For $|x| > 1$, we have $f_{XY}(x, y) = 0$ for all y , and we obviously have

$$f_X(x) = 0.$$

For $|x| \leq 1$, we have $f_{XY}(x, y) = 1/\pi$ for all $|y| < \sqrt{1 - x^2}$ and zero elsewhere. Thus, we have

$$f_X(x) = \int_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} f_{XY}(x, y) dy = \frac{2}{\pi} \sqrt{1-x^2}$$

Totally, we have

$$f_X(x) = \begin{cases} \frac{2}{\pi} \sqrt{1-x^2}, & |x| \leq 1, \\ 0, & \text{elsewhere.} \end{cases}$$

- c. When we have $X = 0$, then Y is uniformly distributed between ± 1 . Thus, we have

$$f_{Y|X}(y|x=0) = \begin{cases} \frac{1}{2}, & |y| \leq 1, \\ 0, & \text{elsewhere.} \end{cases}$$

When we have $X = 0.9$, then Y is uniformly distributed between $\pm \sqrt{1 - 0.9^2} = \pm \sqrt{0.19}$. Thus, we have

$$f_{Y|X}(y|x=0.9) = \begin{cases} \frac{1}{2\sqrt{0.19}}, & |y| \leq \sqrt{0.19}, \\ 0, & \text{elsewhere.} \end{cases}$$

Answer:

a. $B = \frac{1}{\pi}$ b. $f_X(x) = 2B\sqrt{1-x^2}$

c. $f_{Y|X}(y|x=0) = \begin{cases} \frac{1}{2}, & |y| \leq 1, \\ 0, & \text{elsewhere.} \end{cases}$ $f_{Y|X}(y|x=0.9) = \begin{cases} \frac{1}{2\sqrt{0.19}}, & |y| \leq \sqrt{0.19}, \\ 0, & \text{elsewhere.} \end{cases}$

1.7

From the problem formulation, we find $Y_1 = X_1 - X_2$. The expectation is linear, and therefore we have

$$m_{Y_1} = m_{X_1} - m_{X_2} = 1 - 2 = -1.$$

Let A be the matrix relating Y to X . The simplest way to calculate the second moments is probably this:

$$C_Y = AC_X A^T = \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 2 & 1 \\ 1 & 3 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} = \begin{pmatrix} 3 & -1 \\ -1 & 7 \end{pmatrix}.$$

We have thus found that we have $\sigma_{Y_1}^2 = 3$. Alternatively, we can do that using the definition of variance:

$$\begin{aligned} \sigma_{Y_1}^2 &= E\{(Y_1 - m_{Y_1})^2\} = E\{(X_1 - m_{X_1} - X_2 + m_{X_2})^2\} \\ &= E\{(X_1 - m_{X_1})^2\} + E\{(X_2 - m_{X_2})^2\} - 2E\{(X_1 - m_{X_1})(X_2 - m_{X_2})\} \\ &= \sigma_{X_1}^2 + \sigma_{X_2}^2 - 2\lambda_{X_1, X_2} = 2 + 3 - 2 \cdot 1 = 3 \end{aligned}$$

Answer:

$$E\{Y_1\} = \mu_1 - \mu_2 = -1 \quad \text{and} \quad \text{Var}\{Y_1\} = 3$$

1.8

From problem Problem 1.6, we have

$$f_{XY}(x, y) = \begin{cases} B, & x^2 + y^2 \leq 1, \\ 0, & \text{elsewhere.} \end{cases}$$

- a. Based on symmetry, we have $m_X = m_Y = 0$. What is left is then to show that we have $E\{XY\} = m_X m_Y = 0$. We have

$$E\{XY\} = \iint_{-\infty}^{\infty} xy \cdot f_{XY}(x, y) dx dy = \iint_{x^2+y^2 \leq 1} xy \cdot \frac{1}{\pi} dx dy = \frac{1}{\pi} \int_{-1}^1 x \int_{-\sqrt{1-x^2}}^{\sqrt{1-x^2}} y dx dy = 0$$

OK, uncorrelated!

- b. We want to show that we have $f_{XY}(x, y) \neq f_X(x)f_Y(y)$, at least for some pair (x, y) . From the solution to Problem 1.6, we have

$$f_X(x) = \begin{cases} \frac{2}{\pi} \sqrt{1-x^2}, & |x| \leq 1, \\ 0, & \text{elsewhere.} \end{cases}$$

and obviously from the symmetry of $f_{XY}(x, y)$, we also have $f_Y(x) = f_X(x)$. We have

$$f_X(x)f_Y(y) = \begin{cases} \frac{4}{\pi^2} \sqrt{1-x^2} \sqrt{1-y^2}, & |x| \leq 1 \text{ and } |y| \leq 1, \\ 0, & \text{elsewhere.} \end{cases}$$

Obviously, we have

$$f_{XY}(x, y) \neq f_X(x)f_Y(y),$$

and the two variables are definitely dependent.

Answer:

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1.9

The stochastic variable Θ is uniformly distributed over $[0, 2\pi)$. Thus, it has the PDF

$$f_{\Theta}(\theta) = \begin{cases} 1/2\pi, & 0 \leq \theta < 2\pi, \\ 0, & \text{elsewhere.} \end{cases}$$

a.

$$\mathbb{E}\{\cos(\Theta)\} = \int_{-\infty}^{\infty} \cos(\theta) \cdot f_{\Theta}(\theta) d\theta = \frac{1}{2\pi} \int_0^{2\pi} \cos(\theta) d\theta = 0$$

b.

$$\mathbb{E}\{\cos^2(\Theta)\} = \int_{-\infty}^{\infty} \cos^2(\theta) \cdot f_{\Theta}(\theta) d\theta = \frac{1}{2\pi} \int_0^{2\pi} \frac{1}{2} (1 + \cos(2\theta)) d\theta = \frac{1}{2}$$

c.

$$\mathbb{E}\{\cos(\Theta) \sin(\Theta)\} = \int_{-\infty}^{\infty} \cos(\theta) \sin(\theta) \cdot f_{\Theta}(\theta) d\theta = \frac{1}{2\pi} \int_0^{2\pi} \frac{1}{2} \sin(2\theta) d\theta = 0$$

If you cannot remember these trigonometric relations, then Euler can be used to quickly prove them.

Answer:

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