

EE464: HOMEWORK 5 SOLUTIONS

Problem 1 Solution

X : Gaussian, $Y = aX + b$ is a linear combination of gaussian. Therefore Y is also gaussian.

$$\begin{aligned} E[Y] &= aE[X] + b = am + b = m' \\ \text{Var}[Y] &= a^2 \text{Var}[X] = a^2 \sigma^2 = \sigma'^2 \end{aligned}$$

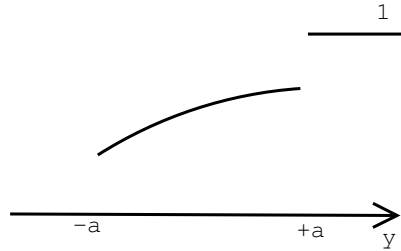
Therefore,

$$a = \sigma' / \sigma, \quad b = m' - am = m' - m\sigma' / \sigma$$

Problem 2 Solution

a)

$$F_Y(x) = \begin{cases} 0 & y < -a \\ F_X(x) & -a \leq y \leq a \\ 1 & y \geq a \end{cases}$$



From sketch of $F_Y(y)$ we see that

$$\begin{aligned} f_Y(y) &= F'_Y(y) = F_X(-a)\delta(y+a) + f_X(y) + (1 - F_Y(a))\delta(y-a) \text{ for } |y| \leq a \\ &= 0 \text{ otherwise} \end{aligned}$$

b)

$f_X(x) = \frac{\beta}{2}e^{-\beta|x|}$ then, taking integration, we obtain

$$F_X(x) = \begin{cases} \frac{1}{2}e^{-\beta x} & x < 0 \\ 1 - \frac{1}{2}e^{-\beta x} & x \geq 0 \end{cases}$$

so,

$$F_Y(y) = \begin{cases} 0 & y < -a \\ \frac{1}{2}e^{-\beta y} & -a \leq y < 0 \\ 1 - \frac{1}{2}e^{-\beta y} & 0 \leq y < a \\ 1 & y \geq a \end{cases}$$

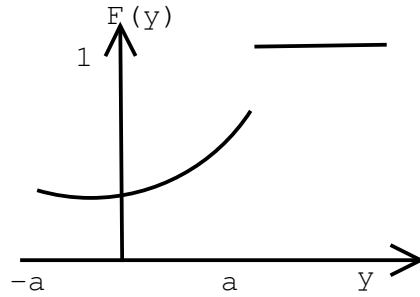
and $f_Y(y) = \frac{1}{2}e^{-\beta a}\delta(y+a) + \frac{\beta}{2}e^{-\beta|y|} + \frac{1}{2}e^{-\beta a}\delta(y-a)$
for $|y| \leq a$ and $f_Y(y) = 0$ elsewhere.

c) For $y < -a$, $F_Y(y) = 0$

For $-a \leq y < a$:

$$F_Y(y) = \int_{-\infty}^a f_X(x) dx + \int_a^y f_X(x) dx$$

For $y \geq a$: $F_Y(y) = 1$

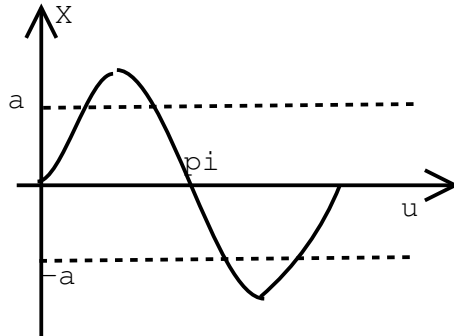


For $-a \leq y \leq a$,

$$f_Y(y) = \int_{-\infty}^a f(x) dx \cdot \delta(y + a) + f_X(y) + \int_a^{\infty} f_X(x) dx \cdot \delta(y - a)$$

d) If $b \leq a$, see Example 3.28 on page 125.

If $b > a$



For $y < -a$, $f_Y(y) = 0$

For $y > a$, $f_Y(y) = 0$

For $-a < y < a$, $f_Y(y) = \frac{1}{\pi\sqrt{1-y^2}}$

$$P[Y = a] = P[Y = -a] = 1 - \frac{2}{\pi} \sin^{-1} \frac{a}{b}$$

Therefore,

$$\begin{aligned} f_Y(y) &= \left(1 - \frac{2}{\pi} \sin^{-1} \frac{a}{b}\right) \delta(y + a) + \frac{1}{\pi\sqrt{1-y^2}} \\ &\quad + \left(1 - \frac{2}{\pi} \sin^{-1} \frac{a}{b}\right) \delta(y - a) \text{ for } |y| \leq a \\ F_Y(y) &= \int_{-\infty}^y f_Y(t) dt \end{aligned}$$

Problem 3 Solutiona) For $y \leq 0$ $P[Y \leq y] = 0$ For $y > 0$ $P[Y \leq y] = P[e^X \leq y] = P[X \leq \ln y] = F_X(\ln y)$

$$F_Y(y) = \begin{cases} 0 & y < 0 \\ F_X(\ln y) & y > 0 \end{cases}$$

For $y > 0$

$$\begin{aligned} f_Y(y) &= \frac{d}{dy} F_Y(y) = F'_X(\ln y) \frac{1}{y} \\ &= \frac{1}{y} f_X(\ln y) \end{aligned}$$

b) If X is a Gaussian random variable, then,

$$f_Y(y) = \begin{cases} 0 & y \leq 0 \\ \frac{e^{-(\ln y - m)^2 / 2\sigma^2}}{y\sqrt{2\pi}\sigma} & y > 0 \end{cases}$$

Problem 4 Solution

a)

$$E[X] = \sum_{k=0}^{\infty} k \frac{\alpha^k}{k!} e^{-\alpha} = \alpha \sum_{k=1}^{\infty} \frac{\alpha^{k-1}}{(k-1)!} e^{-\alpha} = \alpha \sum_{k'=0}^{\infty} \frac{\alpha^{k'}}{k'!} e^{-\alpha} = \alpha$$

$$\begin{aligned} E[X^2] &= \sum_{k=0}^{\infty} k^2 \frac{\alpha^k}{k!} e^{-\alpha} = \alpha \sum_{k=1}^{\infty} k \frac{\alpha^{k-1}}{(k-1)!} e^{-\alpha} \\ &= \alpha \sum_{k'=0}^{\infty} (k'+1) \frac{\alpha^{k'}}{k'!} e^{-\alpha} = \alpha(\alpha+1) \\ \sigma^2 &= E[X^2] - E^2[X] \\ &= \alpha \end{aligned}$$

b) $E[X] = \alpha = \lambda t = \text{arrival rate} * \text{time}$ **Problem 5 Solution**

$$E[X] = \int_{-\infty}^0 x \frac{1}{\pi(1+x^2)} dx + \int_0^{\infty} \frac{x}{\pi(1+x^2)} dx$$

Consider the second term,

$$\int_0^y \frac{x}{\pi(1+x^2)} dx = \frac{1}{2\pi} \ln(1+x^2) \Big|_0^y = \frac{\ln(1+y)}{2\pi} \rightarrow \infty$$

Therefore the integral does not exist, and so $E[X]$ does not exist.**Problem 6 Solution** $X \sim U(0, 1)$

$$\begin{aligned}
F_Y(y) &= P[Y \leq y] = P[-\ln X \leq y] = P[\ln X \geq -y] \\
&= P[X \geq e^{-y}] = 1 - P[X < e^{-y}] = 1 - P[X \leq e^{-y}] \\
&= \begin{cases} 1 - e^{-y} & y > 0 \\ 0 & y \leq 0 \end{cases}
\end{aligned}$$

This is because X is uniformly distributed and $P[X \leq k] = k$ for $0 < k < 1$.
To find density,

$$f_Y(y) = F'(y) = \begin{cases} e^{-y} & y > 0 \\ 0 & y \leq 0 \end{cases}$$

Therefore Y is exponentially distributed.

Problem 7 Solution

$$F_X(x) = \int_0^x e^{-t} dt = \begin{cases} 0 & x < 0 \\ 1 - e^{-x} & x \geq 0 \end{cases}$$

$$\begin{aligned}
F_Y(y) &= P[Y \leq y] = P[e^{-X} \leq y] \\
&= P[-X \leq \ln y] = P[X \geq -\ln y] = 1 - F_X(-\ln y) \\
&= \begin{cases} e^{\ln y} = y & 0 < y < 1 \\ 0 & y \leq 0 \\ 1 & y \geq 1 \end{cases}
\end{aligned}$$

Therefore Y is a uniform r.v.

Problem 8 Solution

$$Y = X^2$$

$$\begin{aligned}
f_Y(y) &= [f_X(\sqrt{y}) + f_X(-\sqrt{y})] \left| \frac{-1}{2\sqrt{y}} \right| \\
&= \frac{1}{\sqrt{2\pi y}} e^{-y/2} \text{ for } y > 0
\end{aligned}$$

This is a chi-square random variable with 1 degree of freedom. See Example 3.25 and 3.26 from the textbook for another way of getting the solution.

Problem 9 Solution

$$\begin{aligned}
F_Y(y) &= P[Y \leq y] = P[|X| \leq y] \\
&= P[-y \leq X \leq y] \\
&= \begin{cases} y & 0 < y < 1 \\ 0 & y \leq 0 \\ 1 & y > 0 \end{cases}
\end{aligned}$$

Therefore Y is a uniform random variable $Y \sim U[0, 1)$.

Problem 10 Solution

a)

$$\begin{aligned} E[X] &= \sum_{k=1}^{\infty} k\beta(1-\beta)^{k-1} = \beta \sum_{k=1}^{\infty} k(1-\beta)^{k-1} \\ &= \beta \left(-\frac{d}{d\beta} \sum_{k=0}^{\infty} (1-\beta)^k \right) \dots \text{Verify!} \\ &= \beta \left(-\frac{d}{d\beta} \left(\frac{1}{\beta} \right) \right) = \frac{1}{\beta} \end{aligned}$$

b)

$$\begin{aligned} E[X^2] &= \sum_{k=1}^{\infty} k^2\beta(1-\beta)^{k-1} = \beta \sum_{k=1}^{\infty} k^2(1-\beta)^{k-1} \\ &= \beta(1-\beta) \sum_{k=1}^{\infty} k^2(1-\beta)^{k-2} \\ &= \beta(1-\beta) \sum_{k=1}^{\infty} k(k-1)(1-\beta)^{k-2} + \sum_{k=1}^{\infty} k\beta(1-\beta)^{k-1} \\ &= \beta(1-\beta) \left(\frac{2}{\beta^3} \right) + \frac{1}{\beta} = \frac{2-\beta}{\beta^2} \end{aligned}$$

Here we used the fact that, $\sum_{k=1}^{\infty} k(k-1)(1-\beta)^{k-2} = \frac{d^2}{d\beta^2} \sum_{k=0}^{\infty} (1-\beta)^k = \frac{d^2}{d\beta^2} \left(\frac{1}{\beta} \right) = 2/\beta^3$

Note that for infinite series, $\frac{d}{dx} \sum f(x) = \sum f'(x)$ is **not** always true. But in this case it has been proved to be true.

$$\begin{aligned} \text{Var}(X) &= E[X^2] - E[X]^2 \\ &= \frac{2-\beta}{\beta^2} - \frac{1}{\beta^2} \\ &= \frac{1-\beta}{\beta^2} \end{aligned}$$

Problem 11 Solution

$$\begin{aligned} E[X] &= \int_0^{\infty} x(2e^{-2x})dx \\ &= \int_0^{\infty} (y/2)(e^{-y})dy \\ &= \frac{1}{2}\Gamma(2) = \frac{1}{2}1! \\ &= \frac{1}{2} \end{aligned}$$

$$\begin{aligned}
 E[X^2] &= \int_0^{\infty} x^2 (2e^{-2x}) dx \\
 &= \frac{1}{4} \int_0^{\infty} y^2 (e^{-y}) dy \\
 &= \frac{1}{4} \Gamma(3) = \frac{1}{4} 2! \\
 &= \frac{1}{2}
 \end{aligned}$$

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

Problem 12 Solution

Using $E[g(X)] = \int g(x)f_X(x)dx$

$$\begin{aligned}
 E[Y] &= E[X^4] = \int_0^1 x^4 3x^2 dx \\
 &= 3[x^7/7]_0^1 \\
 &= \frac{3}{7}
 \end{aligned}$$

Similarly,

$$\begin{aligned}
 E[Y^2] &= E[X^8] = \int_0^1 x^8 3x^2 dx \\
 &= 3[x^{11}/11]_0^1 \\
 &= \frac{3}{11}
 \end{aligned}$$

$$\begin{aligned}
 \text{Var}(Y) &= E[Y^2] - E[Y]^2 \\
 &= \frac{3}{11} - \frac{9}{49} \\
 &= 0.089
 \end{aligned}$$

Problem 13 Solution

a) Using Tchebycheff's inequality,

$$P[|X - \mu| \geq \frac{3}{2}\sigma] \leq \left(\frac{2}{3}\right)^2 = 4/9 = 0.4444$$

b)

$$\begin{aligned}
 P[|X - 1| \geq \frac{3}{2}(\frac{1}{3})] &= P[X - 1 \geq 0.5] + P[1 - X \geq 0.5] = P[X \geq 1.5] + P[X \leq 0.5] \\
 &= \frac{\sqrt{3}}{2}((1 + 1/\sqrt{3}) - 1.5) + \frac{\sqrt{3}}{2}(0.5 - (1 - 1/\sqrt{3})) = 0.067 + 0.067 = 0.134
 \end{aligned}$$

Thus we can see that the inequality is not really tight in this particular case.

Problem 14 Solution

$$\begin{aligned}
 M_X(s) &= E[e^{sX}] \\
 &= \frac{1}{b-a} \int_a^b e^{sx} \cdot 1 dx \\
 &= \frac{1}{b-a} \left. \frac{e^{sx}}{s} \right|_a^b \\
 &= \frac{e^{sb} - e^{sa}}{s(b-a)}
 \end{aligned}$$

This exists when $s \neq 0$.

Problem 15 Solution

$$\begin{aligned}
 M_X(s) &= \sum_{k=1}^{\infty} e^{sk} p(1-p)^{k-1} \\
 &= e^s p \sum_{k=1}^{\infty} e^{s(k-1)} (1-p)^{k-1} \\
 &= e^s p \sum_{k=1}^{\infty} e^{s(k-1)} (1-p)^{k-1} \\
 &= \frac{e^s p}{1 - e^s(1-p)}
 \end{aligned}$$

The condition for convergence of geometric series is $|e^s(1-p)| < 1$. If $s = \alpha + j\beta$, then $|e^s(1-p)| = e^\alpha(1-p) < 1 \Leftrightarrow \text{Re}\{s\} = \alpha < -\ln(1-p)$. Therefore condition for existence of $M_X(s)$ is $\text{Re}\{s\} < -\ln(1-p)$.

$$\begin{aligned}
 M'(s) &= \frac{(1 - e^s(1-p))pe^s + pe^s e^s(1-p)}{(1 - e^s(1-p))^2} \\
 E[X] &= M'(0) = \frac{p^2 + p - p^2}{p^2} \\
 &= \frac{1}{p}
 \end{aligned}$$

Similarly,

$$\begin{aligned}M'(s) &= \frac{(1 - e^s(1 - p))pe^s + pe^s e^s(1 - p)}{(1 - e^s(1 - p))^2} \\ &= \frac{pe^s}{(1 - e^s(1 - p))^2} \\ M''(s) &= \frac{(1 - e^s(1 - p))^2 pe^s + pe^s 2(1 - p)e^s(1 - e^s(1 - p))}{(1 - e^s(1 - p))^4} \\ E[X^2] &= M''(0) = \frac{p^2 p + 2p^2(1 - p)}{p^4} \\ &= \frac{2 - p}{p^2} \\ \text{Var}(X) &= E[X^2] - E[X]^2 \\ &= \frac{2 - p}{p^2} - \frac{1}{p^2} \\ &= \frac{1 - p}{p^2}\end{aligned}$$

TA: Anand Joshi
email: ajoshi@usc.edu