

1. 3.51, p. 180 (Hint: use symmetry where possible.)

The possible values of Y are $-3.5d, -2.5d, -1.5d, -.5d, .5d, 1.5d, 2.5d, 3.5d$. So Y is a discrete random variable. To find the pmf of Y , we need to find the probability of every possible value of Y

(a) The pmf of Y is:

$$p_Y(-3.5d) = P(-\infty \leq X \leq -3d) = \int_{-\infty}^{-3d} f_X(x) dx = \int_{-\infty}^{-3d} \frac{\alpha}{2} e^{-\alpha|x|} dx = \frac{1}{2} e^{-3ad}$$

by symmetry $p_Y(3.5d) = p_Y(-3.5d) = \frac{1}{2} e^{-3ad}$

$$p_Y(-2.5d) = p_Y(2.5d) = P(-3d \leq X \leq -2d) = \int_{-3d}^{-2d} f_X(x) dx = \int_{-3d}^{-2d} \frac{\alpha}{2} e^{-\alpha|x|} dx = \frac{1}{2} (e^{-2ad} - e^{-3ad})$$

$$p_Y(-1.5d) = p_Y(1.5d) = P(-2d \leq X \leq -d) = \int_{-2d}^{-d} f_X(x) dx = \int_{-2d}^{-d} \frac{\alpha}{2} e^{-\alpha|x|} dx = \frac{1}{2} (e^{-ad} - e^{-2ad})$$

$$p_Y(-.5d) = p_Y(.5d) = P(-d \leq X \leq 0) = \int_{-d}^0 f_X(x) dx = \int_{-d}^0 \frac{\alpha}{2} e^{-\alpha|x|} dx = \frac{1}{2} (1 - e^{-ad})$$

and finally $p_Y(y) = 0$ for all other values of y

$$(b) P(|X| > 4d) = 2 \int_{4d}^{\infty} f_X(x) dx = \int_{4d}^{\infty} \frac{\alpha}{2} e^{-\alpha|x|} dx = e^{-4ad}$$

2. 3.53, p. 180

X , the random variable representing the grades, is Gaussian with mean m and standard deviation σ .

Its pdf is $f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-m)^2}{2\sigma^2}}$.

Let $Y = aX + b$. We wish to choose a and b so that Y is Gaussian with mean m' and standard deviation σ' .

Solution 1: $F_Y(y) = \Pr(Y \leq y) = \Pr(aX + b \leq y) = \Pr(X \leq \frac{y-b}{a}) = F_X(\frac{y-b}{a})$.

$$\begin{aligned} f_Y(y) &= \frac{d}{dy} F_Y(y) = \frac{d}{dy} F_X\left(\frac{y-b}{a}\right) = \frac{d}{dx} F_X(x) \Big|_{x=(y-b)/a} \frac{d}{dy} \frac{y-b}{a} = f_X\left(\frac{y-b}{a}\right) \frac{1}{a} \\ &= \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{((y-b)/a - m)^2}{2\sigma^2}} \frac{1}{a} \\ &= \frac{1}{\sqrt{2\pi}a\sigma} e^{-\frac{(y-am-b)^2}{2a^2\sigma^2}}. \end{aligned}$$

We recognize the above as a Gaussian density with mean $am+b$ and standard deviation $a\sigma$.

Equation $m' = am + b$ and $\sigma' = a\sigma$ yields $\mathbf{a} = \frac{\sigma'}{\sigma}$, $\mathbf{b} = m' - am = \mathbf{m}' - \frac{\sigma'}{\sigma} \mathbf{m}$

Solution 2: $m' = E[Y] = E[aX+b] = a E[X] + b$ by linearity of expectation
 $= am + b$

$$(\sigma')^2 = \text{var}(Y) = E (Y-E[Y])^2 = E (aX+b - (a E[X] + b))^2 = E(a(X-E[X]))^2$$

$$= a^2 E (X-E[X])^2 \text{ by linearity of expectation}$$

$$= a^2 \sigma^2$$

Solving for a and b yields, $\mathbf{a} = \frac{\sigma'}{\sigma}$, $\mathbf{b} = \mathbf{m}' - \frac{\sigma'}{\sigma} \mathbf{m}$

3. 3.56, p. 180

$P = R X^2$ where X is Gaussian with $f_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-x^2/2\sigma^2}$.

First we find the cdf:

$$\text{for } y \geq 0, \quad F_P(y) = P(P \leq y) = P(RX^2 \leq y) = P(-\sqrt{y/R} \leq X \leq \sqrt{y/R})$$

$$= F_X(\sqrt{y/R}) - F_X(-\sqrt{y/R})$$

and for $y < 0$, $F_P(y) = 0$

Then we take the derivative of the cdf to obtain the pdf:

for $y \geq 0$,

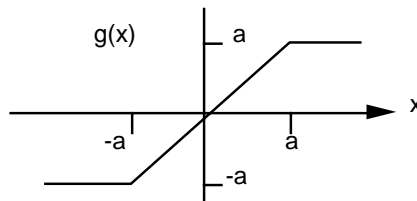
$$\mathbf{f_P(y)} = \frac{d}{dy} F_P(y) = f_X(\sqrt{y/R}) (1/2) (y/R)^{-1/2} (1/R) - f_X(-\sqrt{y/R}) (-1/2) (y/R)^{-1/2} (1/R)$$

$$= \frac{1}{2\sqrt{yR}} (f_X(\sqrt{y/R}) + f_X(-\sqrt{y/R})) = \frac{1}{2\sqrt{yR}} 2 \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{y}{R} \frac{1}{2\sigma^2}\right\}$$

$$= \frac{1}{\sqrt{2\pi\sigma^2 y R}} \exp\left\{-\frac{y}{2\sigma^2 R}\right\}$$

and **for $y < 0$, $f_P(y) = 0$**

4. 3.57 a, p. 180 (Assume X is a continuous random variable. Express your answers in terms of the cdf and/or pdf of X .)



(a) cdf: **for $y < -a$, $F_Y(y) = P(Y \leq y) = 0$,**

for $-a \leq y < a$, $F_Y(y) = P(Y \leq y) = P(g(X) \leq y) = P(X \leq y) = F_X(y)$

for $y \geq a$, $F_Y(y) = P(Y \leq y) = 1$

pdf: **for $y < -a$ or $y > a$, $f_Y(y) = 0$**

$$\begin{aligned} \text{for } -a \leq y \leq a, f_Y(y) &= \frac{d}{dy} F_Y(y) = P(X \leq -a) \delta(y+a) + f_X(y) + P(X \geq a) \delta(y-a) \\ &= F_X(-a) \delta(y+a) + f_X(y) + (1 - F_X(a) + P(X=a)) \delta(y-a) \end{aligned}$$

(c) cdf: for $y < -a$, $F_Y(y) = 0$

$$\text{for } -a \leq y < a, F_Y(y) = F_X(y) = \int_{-\infty}^y \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(z-m)^2}{2\sigma^2}\right\} dz$$

for $y \geq a$, $F_Y(y) = 1$

pdf: for $y < -a$ or $y > a$, $f_Y(y) = 0$

$$\begin{aligned} \text{for } -a \leq y \leq a, f_Y(y) &= \left(\int_{-\infty}^{-a} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(z-m)^2}{2\sigma^2}\right\} dz \right) \delta(y+a) \\ &+ \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(y-m)^2}{2\sigma^2}\right\} + \left(\int_a^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(z-m)^2}{2\sigma^2}\right\} dz \right) \delta(y-a) \end{aligned}$$

5. 3.65, p. 181

This problem is about a discrete random variable with outcomes $\{1, 2, \dots, n\}$ and pmf

$$p_X(i) = \begin{cases} \frac{1}{n}, & i \text{ in } \{1, 2, \dots, n\} \\ 0, & \text{otherwise} \end{cases}$$

The mean

$$E[X] = \sum_{i=1}^n i p_X(i) = \sum_{i=1}^n i \frac{1}{n} = \frac{1}{n} \sum_{i=1}^n i = \frac{1}{n} \frac{n(n+1)}{2} = \frac{n+1}{2}$$

Note that an easy way to derive the fact that $\sum_{i=1}^n i = (n+1)/2$ is that

$$2 \sum_{i=1}^n i = (1 + 2 + \dots + n) + (n + n-1 + \dots + 1) = (n+1) + (n+1) + \dots + (n+1) = n(n+1)$$

To compute the variance we use: $\text{var}(X) = E[X^2] - (E[X])^2$

So we need to find

$$E[X^2] = \sum_{i=1}^n i^2 p_X(i) = \sum_{i=1}^n i^2 \frac{1}{n} = \frac{1}{n} \sum_{i=1}^n i^2 = \frac{1}{n} \frac{n(n+1)(2n+1)}{6} = \frac{(n+1)(2n+1)}{6}$$

Then

$$\text{var}(X) = \frac{(n+1)(2n+1)}{6} - \left(\frac{n+1}{2}\right)^2 = \frac{n^2-1}{12}$$

6. 3.75 (a), p. 182

The net reward is $aX^2 + bX - nd$, where X is a binomial random variable with pmf

$$p_X(x) = \begin{cases} \binom{n}{x} 2^{-n}, & x = 0, 1, \dots, n \\ 0, & \text{else} \end{cases}$$

The expected net reward is

$$E[aX^2 + bX - nd] = a E[X^2] + b E[X] - nd$$

From Table 3.1, $E[X] = n/2$ and $\text{var}(X) = n/4$.

Therefore, $E[X^2] = \text{var}(X) + (E[X])^2 = n^2/4 + n^2/4 = n^2/2$. Finally,

$$E[aX^2 + bX - nd] = a \frac{n^2}{2} + b \frac{n}{2} - nd$$

7. 3.79, p. 182

X is a discrete random variable with pmf $p_X(x) = \begin{cases} \frac{1}{n}, & x = 1, \dots, n \\ 0, & \text{else} \end{cases}$

$$E[Y] = E[K+LX] = K + L E[X] = K + L \frac{n+1}{2} \quad \text{from Problem 3}$$

$$\begin{aligned} \text{var}(Y) &= L^2 \text{var}(X) \quad \text{from Problem 1} \\ &= L^2 \frac{n^2-1}{12} \quad \text{from Problem 3} \end{aligned}$$

8. The "game show problem" is due with the next assignment.