

Berto Federica

Dept: Industrial and Systems Engineering

2.1.8 (a) $X = \{-6, -4, -2, 0, 2, 4, 6\}$

probability mass function (p.m.f.) is:

x_i	-6	-4	-2	0	2	4	6
p_i	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{2}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$

(b) cumulative distribution function:

$$F(x) = \begin{cases} 0 & , x < -6 \\ \frac{1}{8} & , -6 \leq x < -4 \\ \frac{2}{8} & , -4 \leq x < -2 \\ \frac{3}{8} & , -2 \leq x < 0 \\ \frac{5}{8} & , 0 \leq x < 2 \\ \frac{6}{8} & , 2 \leq x < 4 \\ \frac{7}{8} & , 4 \leq x < 6 \\ 1 & , x \geq 6 \end{cases}$$

(c) the most likely value of X^* is the one for which $P(X^*) \geq P(x_i) \forall x_i$
 \rightarrow so the value $X^* = 0$.

2.2.10

$$F(x) = \begin{cases} \frac{0.25}{500} x & 0 \leq x < 500 \\ \frac{0.25}{500} x + 0.5 & 500 \leq x < 1000 \end{cases}$$

so, (a) $P(X \leq 200) = F(200) = 200 \cdot \frac{0.25}{500} = [0.1]$

(b) $F(700) - F(400) = 700 \cdot \frac{0.25}{500} + 0.5 - 400 \cdot \frac{0.25}{500} = [0.65]$

The probability on the second range appears much higher because it contains the probability of winning 500\$, which is 0.5.

2.3.14

$$\begin{cases} F(0) = A + \frac{1}{125} B = 0 \\ F(50) = A + \frac{1}{55^3} B = 1 \end{cases} \Rightarrow \text{we obtain: } \begin{cases} A = 1.00075 \\ B = -125.083 \end{cases}$$

$\Rightarrow F(r) = 1.00075 - \frac{125.083}{(r+5)^3}$

$f(r) = \frac{dF(r)}{dr} = \frac{375.36}{(r+5)^4}$ with $r \in [0, 50]$

$$\text{So } E(r) = \int_{-\infty}^{\infty} F(x) \cdot x \, dx = \int_0^{50} \frac{375.36}{(r+5)^4} \, dr \approx [2.44]$$

Median derivation is obtained by:

$$F(r) = 0.5 \rightarrow 1.00075 - \frac{125.093}{(r+5)^3} = 0.5 \Rightarrow [r \approx 1.3]$$

[2.4.8] $\int_0^{10} f(\theta) \, d\theta = 1$ because it is a probability distribution

$$\Rightarrow \int_0^{10} A(e^{10-\theta} - 1) \, d\theta = 1 \Rightarrow A = \frac{1}{e^{10} - 11}$$

② $\sigma^2 = E(x^2) - \mu^2$, so:

$$E(x^2) = \int_0^{10} \frac{(e^{10-\theta} - 1)}{e^{10} - 11} \theta^2 \, d\theta \approx 1.98$$

$$\mu^2 = \left[\int_0^{10} \frac{(e^{10-\theta} - 1)}{e^{10} - 11} \theta \, d\theta \right]^2 = 0.99773^2 \approx 0.995$$

$$\text{So: } \sigma^2 = 1.98 - 0.995 = [0.985]$$

③ $\sigma = \sqrt{\sigma^2} = \sqrt{0.985} = [0.992]$

④ Upper quantile: $F(\theta) = 0.75$, so:

$$\int_0^{\theta} \frac{e^{10-\theta} - 1}{e^{10} - 11} \, d\theta = 0.75 \Rightarrow [\theta \approx 1.385]$$

Lower quantile: $F(\theta) = 0.25$:

$$\int_0^{\theta} \frac{e^{10-\theta} - 1}{e^{10} - 11} \, d\theta = 0.25 \Rightarrow [\theta \approx 0.288]$$

⑤ IQR (Interquartile range) = $Q_3 - Q_1 = 1.385 - 0.288 = [1.097]$

[2.5.2] ② Probability mass function:

$x_i \backslash y_j$	0	1	2	3
0	$\frac{1}{16}$	$\frac{1}{16}$	0	0
1	$\frac{1}{16}$	$\frac{3}{16}$	$\frac{2}{16}$	0
2	0	$\frac{2}{16}$	$\frac{3}{16}$	$\frac{1}{16}$
3	0	0	$\frac{1}{16}$	$\frac{1}{16}$

③ Marginal Probability ^{m.p.} of X:

x_i	0	1	2	3
p_i	$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$

Marginal Probability m.p. of Y:

y_j	0	1	2	3
p_j	$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$

④ Independence if $f(x, y) = f_x(x) \cdot f_y(y)$

But i.e. $\rightarrow f(0, 0) = \frac{1}{16} \neq f_x(0) \cdot f_y(0) = \frac{1}{8} \cdot \frac{1}{8} = \frac{1}{64}$

Since $f(0, 0) \neq f_x(0) \cdot f_y(0)$, then X, Y are NOT independent.

$$d) \bullet E(X) = \sum_{i=0}^3 p_i x_i = \left[\frac{3}{2} \right] = \mu_x$$

$$E(X^2) = \sum_{i=0}^3 p_i x_i^2 = 3$$

$$\bullet \sigma_x^2 = E(X^2) - \mu_x^2 = 3 - \left(\frac{3}{2}\right)^2 = \left[\frac{3}{4}\right]$$

$$\bullet E(Y) = \sum_{i=0}^3 p_i y_i = \left[\frac{3}{2} \right] = \mu_y$$

$$E(Y^2) = \sum_{i=0}^3 p_i y_i^2 = 3$$

$$\bullet \sigma_y^2 = E(Y^2) - \mu_y^2 = 3 - \left(\frac{3}{2}\right)^2 = \left[\frac{3}{4}\right]$$

$$e) \text{Cov}(X, Y) = E(XY) - E(X)E(Y)$$

$$E(XY) = \sum_{i=0}^3 \sum_{j=0}^3 p_{ij} \cdot x_i \cdot y_j = \frac{44}{16}$$

$$\Rightarrow \text{Cov}(X, Y) = \frac{44}{16} - \frac{3}{2} \cdot \frac{3}{2} = \left[\frac{1}{2}\right]$$

$$f) P(X=0 | Y=1) = \frac{p_{01}}{p_{+1}} = \frac{1}{16} / \frac{6}{16} = \left[\frac{1}{6}\right]$$

$$P(X=1 | Y=1) = \frac{p_{11}}{p_{+1}} = \frac{3}{16} / \frac{6}{16} = \left[\frac{1}{2}\right]$$

$$P(X=2 | Y=1) = \frac{p_{21}}{p_{+1}} = \frac{1}{8} / \frac{6}{16} = \left[\frac{1}{3}\right]$$

$$P(X=3 | Y=1) = \frac{p_{31}}{p_{+1}} = 0 / \frac{6}{16} = [0]$$

$$E(X | Y=1) = 0 \cdot \frac{1}{6} + 1 \cdot \frac{1}{2} + 2 \cdot \frac{1}{3} = \left[\frac{7}{6}\right]$$

$$E(X^2 | Y=1) = 0^2 \cdot \frac{1}{6} + 1^2 \cdot \frac{1}{2} + 2^2 \cdot \frac{1}{3} = \left[\frac{11}{6}\right]$$

$$\Rightarrow \sigma^2(X | Y=1) = E(X^2 | Y=1) - [E(X | Y=1)]^2 = \left[\frac{17}{36}\right]$$

2.6.4 Since they are independent,

$$\sigma^2(0.5X_1 + 0.3X_2 + 0.2X_3) = \sigma^2(0.5X_1) + \sigma^2(0.3X_2) + \sigma^2(0.2X_3)$$

$$\Rightarrow \sigma = \sqrt{0.5^2 \cdot 1.2^2 + 0.3^2 \cdot 2.4^2 + 0.2^2 \cdot 3.12^2} = [1.24]$$

2.6.16

$$a) E(A) = E\left(\frac{X_1 + X_2}{2}\right) = \frac{1}{2} E(X_1) + \frac{1}{2} E(X_2) = \frac{1}{2} \omega + \frac{1}{2} \omega = [\omega]$$

$$\sigma(A) = \sqrt{\sigma^2\left(\frac{X_1}{2}\right) + \sigma^2\left(\frac{X_2}{2}\right)} = \sqrt{\frac{1}{4} \cdot 3^2 + \frac{1}{4} \cdot 4^2} = \left[\frac{5}{2}\right]$$

$$b) \sigma^2(B) = \delta^2 \sigma^2(X_1) + (1-\delta)^2 \sigma^2(X_2) = 25\delta^2 - 32\delta + 16$$

$$\text{we minimize } f(\delta) \rightarrow f'(\delta) = 0 = 50\delta - 32 \rightarrow \left[\delta = \frac{16}{25}\right]$$

$$\text{So, } \sigma^2(B) \text{ given } \delta = \frac{16}{25} = \frac{144}{25}; \text{ The minimum standard deviation will be:}$$

$$\sigma(B) \Big|_{\delta = \frac{16}{25}} = \left[\frac{12}{5}\right]$$

9.6

$$\textcircled{a} \begin{cases} E(Y) = 100 = E(d \cdot c) = dE(X) + c = 250d + c \\ \sigma^2(Y) = \sigma^2(dX + c) = d^2 \sigma^2(X) = 16d^2 = 1 \end{cases}$$

$$\Rightarrow \text{we obtain: } \begin{cases} d = 0.25 \\ c = 37.5 \end{cases}$$

$$\textcircled{b} E(\underbrace{X + X + \dots + X}_{10}) = \underbrace{E(X) + E(X) + \dots + E(X)}_{10} = 10E(X) = 2500$$

$$\sigma^2(\underbrace{X + X + \dots + X}_{10}) = \underbrace{\sigma^2(X) + \sigma^2(X) + \dots + \sigma^2(X)}_{10} = 10\sigma^2(X) = 160$$

$$\text{Thus, } \sigma(10 \times 1) = \sqrt{10 \sigma^2(X)} = \sqrt{10 \cdot 16} = [4\sqrt{10}]$$

$$F(x) = AX$$

9.16

$$\textcircled{a} \int_5^6 AX dx = 1 \rightarrow A \left[\frac{x^2}{2} \right]_5^6 = 1; \left[A = \frac{2}{11} \right]$$

$$\textcircled{b} \text{c.d.f.} = \int_5^x \frac{2}{11} x dx = \left[\frac{x^2 - 25}{11} = F(x) \right]$$

$$\textcircled{c} E(X) = \int_5^6 \frac{2x}{11} \cdot x dx = \frac{182}{33} = 4$$

$$\textcircled{d} \sigma = \sqrt{E(X^2) - \mu^2}$$

$$E(X^2) = \int_5^6 \frac{2}{11} x \cdot x^2 dx = \frac{671}{22}$$

$$\text{So, } \sigma = \sqrt{\frac{671}{22} - \left(\frac{182}{33}\right)^2} = \sqrt{\frac{875}{1089}} \approx [0.289]$$

9.24

① Marginal p.d.f. of X:

$$f_X(x) = \int_0^2 4x(2-y) dy = [2x], \quad x \in [0, 1]$$

② Marginal p.d.f. of Y:

$$f_Y(y) = \int_0^1 4x(2-y) dx = 2(2-y)$$

Independence: if $f(x, y) = f_X(x) \cdot f_Y(y)$, so:

$$4x(2-y) = 2x \cdot 2(2-y) = 4x(2-y) \quad \checkmark$$

They are independent.

③ $\text{cov}(X, Y) = 0$, because Independence implies zero covariance.

④ $F_{X|Y=1.5}(x) = f_X(x) = 2x$, because x and y are independent.