

Statistics - Assignment - 3

5.34

$$E(X) = \left[-\frac{1}{8} - \frac{0.5}{4} + \frac{0.5}{2} + \frac{1}{8} \right]$$

$$= \frac{-1 - 1 + 2 + 1}{8} = \frac{1}{8} = 0.125$$

$$E(Y) = \frac{-2}{8} - \frac{1}{4} + \frac{1}{2} + \frac{2}{8}$$

$$= -\frac{1}{4} - \frac{1}{4} + \frac{1}{2} + \frac{1}{4} = \frac{1}{2} - \frac{1}{4} = \frac{2-1}{4} = \frac{1}{4}$$

$$= 0.25$$

$$E(XY) = \frac{2}{8} + \frac{0.5}{4} + \frac{0.5}{2} + \frac{2}{8} = \frac{4}{8} + 0.5 \left(\frac{1}{4} + \frac{1}{2} \right)$$

$$= \frac{1}{2} + 0.5 \left(\frac{1+2}{4} \right) = \frac{1}{2} + 0.5 \left(\frac{3}{4} \right)$$

$$= 0.875$$

$$E(X^2) = \frac{1}{8} + \frac{0.25}{4} + \frac{0.25}{2} + \frac{1}{8} = 0.4375$$

$$E(Y^2) = \frac{4}{8} + \frac{1}{4} + \frac{1}{2} + \frac{2}{8} = 1.75$$

$$V(X) = E(X^2) - [E(X)]^2$$

$$= 0.4375 - 0.015625 = 0.421875$$

$$V(Y) = E(Y^2) - [E(Y)]^2$$

$$= 1.75 - 0.0625 = 1.6875$$

⇒ Covariance:

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

$$= 0.875 - (0.125)(0.25) = 0.875 - 0.03125$$

$$= 0.84375$$

⇒ Correlation coefficient:

$$P_{xy} = \frac{\text{Cov}(X, Y)}{\sqrt{V(X)V(Y)}} = \frac{\sigma_{xy}}{\sigma_x \sigma_y}$$

$$= \frac{0.84375}{\sqrt{(0.421875)(1.6875)}} = \frac{0.84375}{\sqrt{0.421875 \times 1.6875}}$$

$$= \frac{0.84375}{0.84375} = 1$$

5.65

$$\mu = 2 \quad \sigma^2 = 0.1$$

$$E(X_1) = 2$$

$$V(X_1) = 0.1$$

$$E(X_2) = 2$$

$$V(X_2) = 0.1$$

$$E(\text{Total}) = E(X_1) + E(X_2)$$

$$= 2 + 2 = 4$$

$$\sigma(\text{Total}) = \sqrt{(\sigma(X_1))^2 + (\sigma(X_2))^2} = \sqrt{0.1^2 + 0.1^2}$$

$$= \sqrt{0.01 + 0.01} = \sqrt{0.02} = 0.141421$$

a) Mean of the total thickness of two halves is 4.

Variance of the total thickness of is
(Standard Deviation) = 0.141421.

$$b) P(T > 4.3) = P\left(\frac{T - E(\text{Total})}{\sigma(\text{Total})} > \frac{4.3 - 4}{0.141421}\right)$$

$$= P(Z > 2.12132) = 1 - P(Z \leq 2.12132)$$

$$= 1 - 0.983414 = 0.016586$$

7.7

$$\mu = 2500$$

$$\sigma = 50$$

$$n = 5$$

$$\sigma_x = \frac{\sigma}{\sqrt{n}} = \frac{50}{\sqrt{5}} = \frac{50}{2.236}$$

$$\sigma_x = \frac{\sigma}{\sqrt{n}} = \frac{50}{\sqrt{5}} = \frac{50}{2.236} = 22.36 \approx 22.421$$

$$P(2499 < X < 2510) = P\left(\frac{2499 - 2500}{22.421} < \frac{X - 2500}{22.421}\right)$$

$$< \frac{2510 - 2500}{22.421})$$

$$= P\left(\frac{-1}{22.421} < Z < \frac{10}{22.421}\right) = P(-0.0446 < Z$$

$$< 0.446)$$

$$= \Phi(0.446) - \Phi(-0.0446)$$

$$= 0.673645 - 0.480061 = 0.193584 \approx 0.193$$

7.36 a) Prove $\hat{p} = \frac{x}{n}$ is an unbiased estimator of p .

Assuming an event K indicates k^{th} event as either success (1) or failure (0).

$$\begin{aligned} \mu_k &= 1(p) + 0(1-p) \\ &= p \end{aligned}$$

Total number of success X is the sum of all the events of E_k .

$$E(X) = \mu_1 + \mu_2 + \dots + \mu_k + \dots + \mu_n$$

$$= p + p + \dots + p = np$$

$$E(\hat{p}) = E\left(\frac{X}{n}\right) = \frac{1}{n} E(X) = \frac{1}{n} (np) = p$$

$\therefore \hat{p} = \frac{x}{n}$ is an unbiased estimator of p .

D) Variance of each event E_k is

$$V(E_i) = E(G_i^2) - [E(G_i)]^2$$

Assume $E(G_i^2) = 1^2(p) + 0^2(1-p) = p$

$$V(G_i) = p - p^2 = p(1-p)$$

Variance of $X = V(G_1) + V(G_2) + V(G_3) + \dots + V(G_n)$

$$= np(1-p)$$

Given $\hat{p} = \frac{x}{n} \Rightarrow V(\hat{p}) = V\left(\frac{x}{n}\right) = \frac{1}{n^2} np(1-p)$

Standard Error:

$$S.E. \text{ of } \hat{p} = \sqrt{V(\hat{p})} = \sqrt{\frac{p(1-p)}{n}}$$

7.24

$$E(\hat{\theta}_1) = E\left[\frac{x_1 + x_2}{2}\right] = \frac{1}{2} [E(x_1) + E(x_2)]$$

$$E(\hat{\theta}_1) = \frac{1}{2} (\mu + \mu) = \mu$$

$$E(\hat{\theta}_2) = E\left[\frac{x_1 + 3x_2}{4}\right] = \frac{1}{4} [E(x_1) + 3E(x_2)]$$

$$E(\hat{\theta}_2) = \frac{\mu}{4} + \frac{3\mu}{4} = \mu$$

So, both $E(\hat{\theta}_1)$ and $E(\hat{\theta}_2)$ are unbiased estimator of μ .

Variance of estimators:

$$V(\hat{\theta}_1) = V\left[\frac{x_1 + x_2}{2}\right] = \frac{1}{4}\sigma^2 + \frac{1}{4}\sigma^2$$

$$= \frac{2}{4}\sigma^2 = \frac{1}{2}\sigma^2$$

$$V(\hat{\theta}_2) = V\left[\frac{x_1 + 3x_2}{4}\right] = \frac{1}{16}\sigma^2 + \frac{9}{16}\sigma^2$$

$$= \frac{10}{16}\sigma^2 = \frac{5}{8}\sigma^2$$

$$V(\hat{\theta}_1) = \frac{1}{2}\sigma^2, \quad V(\hat{\theta}_2) = \frac{5}{8}\sigma^2$$