

326 However k d.

1 A.  $E[Y_1] = \mu$ , unbiased  
 $MSE(Y_1) = V(Y_1) + B(Y_1)^2 = \sigma^2$

B.  $E[Y_1 + Y_2] = 2\mu$ , Biased  
 $B(Y_1 + Y_2) = 2\mu - \mu = \mu$

C.  $E\left[\frac{Y_1 + 2Y_2 + 2Y_3 + Y_4}{6}\right] = \frac{\mu + 2\mu + 2\mu + \mu}{6} = \mu$ , unbiased  
 $V\left(\frac{Y_1 + 2Y_2 + 2Y_3 + Y_4}{6}\right) = \frac{\sigma^2 + 4\sigma^2 + 4\sigma^2 + \sigma^2}{36} = \frac{5\sigma^2}{18}$   
 $MSE\left(\frac{Y_1 + 2Y_2 + 2Y_3 + Y_4}{6}\right) = V(\ ) + B(\ ) = \frac{5\sigma^2}{18}$

D.  $E[\bar{Y}] = \mu$ ,  $V[\bar{Y}] = \frac{\sigma^2}{4}$  as  $n=4$ .  
 $MSE(\bar{Y}) = V[\bar{Y}] = \frac{\sigma^2}{4}$  lowest.



2. Have  $MSE(\hat{\beta}_1) = \frac{p - p^2}{n} \Rightarrow \frac{p - p^2}{15}$

$$MSE(\hat{\beta}_0) = \frac{1 + (1-4)p + (4-n)p^2}{(n+d)^2} \Rightarrow \frac{1 + 11p - 11p^2}{289}$$

If  $\hat{\beta}_0$  is better, then  $MSE(\hat{\beta}_1) - MSE(\hat{\beta}_0) > 0$ .

$$D(p) = MSE(\hat{\beta}_1) - MSE(\hat{\beta}_0) = \frac{-15 + 124p - 124p^2}{4331}$$

$$D(p) = 0 \text{ when } p = \frac{31 \pm 4\sqrt{31}}{62}$$

Since  $D(1/2) = 44 > 0$ ,  $\hat{\beta}_0$  is the better estimator when

$$\frac{31 - 4\sqrt{31}}{62} < p < \frac{31 + 4\sqrt{31}}{62}$$



$$3. a) E(\bar{Y}) = E(Y) = \int_0^{\theta} y \cdot \frac{3y^2}{\theta^3} dy = \frac{3}{\theta^3} \int_0^{\theta} y^3 dy = \frac{3}{4} \theta.$$

$\neq \theta \Rightarrow \text{biased}$

$$b) \text{ let } \tilde{\theta}_1 = \frac{4}{3} \hat{\theta}_1 = \frac{4}{3} \bar{Y}.$$

$$c) \text{ Since } B(\tilde{\theta}_1) = 0, \text{MSE}(\tilde{\theta}_1) = V(\tilde{\theta}_1)$$

$$V(\tilde{\theta}_1) = V\left(\frac{4}{3} \bar{Y}\right) = \frac{16}{9} V(\bar{Y}) = \frac{16}{9} \cdot \frac{V(Y)}{n}.$$

For  $V(Y)$ , use  $E(Y^2) - E(Y)^2$ .

$$E(Y^2) = \int_0^{\theta} y^2 \cdot \frac{3y^2}{\theta^3} dy = \frac{3}{\theta^3} \int_0^{\theta} y^4 dy = \frac{3\theta^2}{5}.$$

$$V(Y) = \frac{3\theta^2}{5} - \left(\frac{3}{4}\theta\right)^2 = \frac{3}{20}\theta^2.$$

$$\text{So } \text{MSE}(\tilde{\theta}_1) = \frac{16}{9} \cdot \frac{3\theta^2/20}{n} = \frac{\theta^2}{15n}$$

$$d) \text{ For max order stat, we have } P(Y_{(n)} \leq y) = [F(y)]^n \quad \text{w/ pdf}$$

$$m(u) = n [F(u)]^{n-1} \cdot f(u).$$

$$\text{for } F(u) = \int_0^u \frac{3y^2}{\theta^3} dy = \left(\frac{y}{\theta}\right)^3$$

$$\text{and } m(u) = n \left(\frac{y}{\theta}\right)^{3(n-1)} \cdot \frac{3y^2}{\theta^3} = \frac{3ny^{3n-1}}{\theta^{3n}} \quad 0 < u < \theta$$



$$\begin{aligned} \text{Then } E(\hat{\theta}_0) &= \int_0^\theta u \cdot \frac{3n u^{3n-1}}{\theta^{3n}} du \\ &= \frac{3n}{\theta^{3n}} \int_0^\theta u^{3n} du = \frac{3n}{3n+1} \theta \neq \theta. \end{aligned}$$

$$e) \tilde{\theta}_0 = \frac{3n+1}{3n} \hat{\theta}_0.$$

$$f) \text{ As before, } \text{MSE}(\tilde{\theta}_0) = V(\tilde{\theta}_0) = \left(\frac{3n+1}{3n}\right)^2 V(\hat{\theta}_0)$$

$$\text{For } V(\hat{\theta}_0), E(u^2) = \int_0^\theta u^2 \cdot \frac{3n u^{3n-1}}{\theta^{3n}} du = \frac{3n}{3n+2} \theta^2.$$

$$\begin{aligned} \text{So } \text{MSE}(\tilde{\theta}_0) &= \left(\frac{3n+1}{3n}\right)^2 \left[ \frac{3n}{3n+2} \theta^2 - \left(\frac{3n}{3n+1} \theta\right)^2 \right] \\ &= \theta^2 \left[ \frac{(3n+1)^2}{3n(3n+2)} - 1 \right] \\ &= \frac{\theta^2}{3n(3n+2)} \end{aligned}$$

Note  $\tilde{\theta}_0$  is a better estimator than  $\tilde{\theta}_1$  when  $n > 1$ .



4.  $p_1$  : proportion of graduates who are firstborn

$p_0$  : proportion of non-graduates who are firstborn.

$$p_1 \text{ estimated by } \hat{p}_1 = \frac{126}{180} = \frac{70}{100}$$

$$\text{Note } n_1 = 180 \text{ and } \sigma_{\hat{p}_1}^2 = \frac{pq}{n_1} \approx \frac{(0.7)(0.3)}{180}$$

$$p_0 \text{ estimated by } \hat{p}_0 = \frac{54}{100}$$

$$\text{Note } n_0 = 100 \text{ and } \sigma_{\hat{p}_0}^2 \approx \frac{(0.54)(0.46)}{100}$$

So  $p_1 - p_0$  estimated by

$$\hat{p}_1 - \hat{p}_0 \pm \sqrt{\sigma_{\hat{p}_1}^2 + \sigma_{\hat{p}_0}^2}$$

$$(0.7 - 0.54) \pm 2(0.0604)$$

$$0.16 \pm 0.1208 \text{ or } (0.04, 0.28)$$



$$\begin{aligned}
 \text{Ex. a) } E(X) &= aE(\hat{\theta}_1) + bE(\hat{\theta}_2) + cE(\hat{\theta}_3) \\
 &= a\theta + b\theta + c\theta \\
 &= (a+b+c)\theta \\
 &= \theta.
 \end{aligned}$$

$$\begin{aligned}
 \text{b) } V(X) &= a^2 V(\hat{\theta}_1) + b^2 V(\hat{\theta}_2) + c^2 V(\hat{\theta}_3) \text{ since indy.} \\
 &= a^2(1+1) + b^2(1+2) + c^2(1+3) \\
 &= 2a^2 + 3b^2 + 4c^2.
 \end{aligned}$$

$$\begin{aligned}
 \text{Minimize } V(X) &= 2a^2 + 3b^2 + 4c^2 \\
 \text{Subject to: } &a + b + c = 1.
 \end{aligned}$$

$$\nabla V = \langle 4a, 6b, 8c \rangle$$

$$\text{Lagrange: } \nabla V \parallel \langle 1, 1, 1 \rangle$$

$$\text{i.e. } 4a = \lambda \quad 4a = 6b \Rightarrow b = \frac{2}{3}a$$

$$\begin{aligned}
 6b &= \lambda \\
 8c &= \lambda
 \end{aligned}
 \quad \left. \begin{array}{l} \\ \end{array} \right\} 4a = 8c \Rightarrow c = \frac{1}{2}a.$$

$$\text{Then } a + \frac{2}{3}a + \frac{1}{2}a = 1, \quad 6a + 4a + 3a = 6, \quad a = \frac{6}{13}$$

$$b = \frac{4}{13}, \quad c = \frac{3}{13}.$$