

Homework 10

● Graded

Student

Jacob Hauptman

Total Points

23 / 24 pts

Question 1

(no title)

5 / 6 pts

– 0 pts Correct

✓ – 1 pt small mistake / missing a small step

– 2 pts missing critical steps

– 3 pts some steps are correct and some are wrong

– 6 pts wrong

1 ?

Question 2

(no title)

6 / 6 pts

✓ – 0 pts Correct

– 1 pt wrong $\text{Var}(X)$ and $\text{Var}(Y)$ or other mistake in b

– 1 pt missing a small step

– 1 pt small mistake

– 1 pt c. incomplete

– 2 pts c. missing

Question 3

(no title)

6 / 6 pts

✓ – 0 pts Correct

– 1 pt small mistake

– 3 pts missing critical steps

– 6 pts Missing / wrong

Question 4

(no title)

6 / 6 pts

✓ - 0 pts Correct

- 1 pt small mistake
- 1 pt b. wrong
- 2 pts missing part b
- 1 pt c. wrong
- 2 pts missing part c
- 6 pts missing

Questions assigned to the following page: [2](#) and [1](#)

1)

$$\begin{aligned}
 E(\hat{p}) &= E\left(\frac{r-1}{x+r-1}\right) = \sum_{x=0}^{\infty} \frac{r-1}{x+r-1} \binom{x+r-1}{x} p^r (1-p)^x = \sum_{x=0}^{\infty} \frac{r-1}{x+r-1} \left[\frac{(x+r-1)!}{x!(r-1)!} \right] p^r (1-p)^x \\
 &= \sum_{x=0}^{\infty} \frac{r-1}{x+r-1} \left[\frac{(x+r-1)(x+r-2)!}{x!(r-1)(r-2)!} \right] p^r (1-p)^x = \sum_{x=0}^{\infty} \frac{(x+r-2)!}{x!(r-2)!} p^r (1-p)^x \\
 &= \sum_{x=0}^{\infty} \binom{x+r-2}{x} p^r (1-p)^x \quad \text{Let } r=k+1, \text{ so} \\
 &= \sum_{x=0}^{\infty} \binom{x+k-1}{x} p^{k+1} (1-p)^x \cdot \frac{(k+1)(1-p)}{p} = \frac{(r-1+1)(1-p)}{p} = \frac{r(1-p)}{p} \quad \square
 \end{aligned}$$

2)

a $B_{ias, \mu}(\hat{\mu}) = E(\hat{\mu}) - \mu = E(\delta \bar{X} + (1-\delta)\bar{Y}) - \mu = E(\delta \bar{X}) + E(\bar{Y}) - E(\delta \bar{Y}) - \mu$

$$= \delta \mu + \mu - \delta \mu - \mu = 0$$

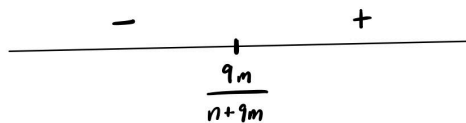
b $V(\hat{\mu}) = \delta^2 V(\bar{X}) + (1-\delta)^2 V(\bar{Y}) = \delta^2 \frac{\sigma^2}{m} + (1-\delta)^2 \frac{q\sigma^2}{n}$

c $\frac{d}{d\delta} \left[\delta^2 \frac{\sigma^2}{m} + (1-\delta)^2 \frac{q\sigma^2}{n} \right] = 2\delta \frac{\sigma^2}{m} - 2(1-\delta) \frac{q\sigma^2}{n}$

Set to 0.

$$2\delta \frac{\sigma^2}{m} = 2(1-\delta) \frac{q\sigma^2}{n}$$

$$\frac{\delta}{m} = \frac{q-1\delta}{n} \Rightarrow n\delta = qm - qm\delta \Rightarrow n\delta + qm\delta = qm \Rightarrow \delta = \frac{qm}{n+qm}$$



So it is a minimum.

Questions assigned to the following page: [3](#) and [4](#)

3)

• a $(X_i - \bar{X})^2 = X_i^2 - 2X_i\bar{X} + \bar{X}^2$

$$E(\sum X_i^2 - 2X_i\bar{X} + \bar{X}^2) = E(\sum X_i^2 - 2\bar{X}\sum X_i + \bar{X}^2 \sum 1) = E(\sum X_i^2 - 2\bar{X}\sum X_i + n\bar{X}^2)$$

$$= E(\sum X_i^2 - 2\bar{X}(n\bar{X}) + n\bar{X}^2) = E(\sum X_i^2 - 2n\bar{X}^2 + n\bar{X}^2) = E(\sum X_i^2 - n\bar{X}^2)$$

$$= \sum E(X_i^2) - E(n\bar{X}^2)$$

$$V(X) = E(X^2) - E(X)^2, \text{ so } E(X^2) = \sigma^2 + \mu^2 \text{ and } E(\bar{X}^2) = \frac{\sigma^2}{n} + \mu^2$$

$$= \sum (\sigma^2 + \mu^2) - n(\frac{\sigma^2}{n} + \mu^2) = n\sigma^2 + n\mu^2 - \sigma^2 - n\mu^2 = n\sigma^2 - \sigma^2 = \sigma^2(n-1)$$

• b $E(S_{xx}) = \sigma^2(n-1) \Rightarrow \sigma^2 = \frac{E(S_{xx})}{n-1}$

Therefore $\frac{S_{xx}}{n-1}$ is an unbiased estimator for σ^2 .

$$\frac{S_{xx}}{n-1} = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1} \text{ as well.}$$

4)

• a $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i, \mu = \ell \text{ so } E(\bar{X}) = E(\frac{1}{n} \sum_{i=1}^n X_i) = \frac{1}{n} \sum_{i=1}^n E(X_i) = \frac{1}{n} \sum_{i=1}^n \mu = \mu = \ell \text{ so } \bar{X} \text{ is an unbiased estimator for } \ell \text{ since } E(\bar{X}) = \mu = \ell. \text{ The estimator equals the true value}$

• b $\text{Bias}_{\ell^2}(\bar{X}^2) = E(\bar{X}^2) - E(X^2) = V(\bar{X}) + E(\bar{X})^2 - \ell^2 = \frac{\sigma^2}{n} + \ell^2 - \ell^2 = \frac{\sigma^2}{n} \text{ so}$

\bar{X} is a bias estimator for ℓ^2 and over estimates by $\frac{\sigma^2}{n}$

• c $\text{Bias}_{\ell^2}(\bar{X}^2 - kS^2) = E(\bar{X}^2) - kE(S^2) - \ell^2 = \frac{\sigma^2}{n} + \ell^2 - k\sigma^2 - \ell^2 = \frac{\sigma^2}{n} - k\sigma^2$

$$\frac{\sigma^2}{n} = k\sigma^2 \Rightarrow \frac{1}{n} = k$$