

a.1 1.8, 9, 13

b.2 20, 21, 29, 32

A

1. a. Estimator = \bar{x}

estimate = $\bar{x} = \sum x_i / n = \frac{219.8}{27} = 8.14$

b. Estimator = \hat{x}

estimate = $\hat{x} = 0.77$

c. Estimator = s

estimate = $s = \sqrt{\frac{\sum x_i^2 - (\sum x_i)^2 / n}{n-1}} = 1.66$

d. $p = \frac{4}{27} = 0.148$

e. Estimator = $\frac{s}{\bar{x}}$

estimate = $\frac{s}{\bar{x}} = \frac{1.66}{8.14} = 0.204$

8.

a. $p = \frac{80-12}{80} = 0.85$

b. ~~$p = p \cdot p = 0.7225$~~

$p = 0.85 \times \frac{67}{79}$
 $= 0.72$

9. a. Estimator ~~$\hat{\theta}$~~ : \bar{x}

$$E(\bar{x}) = \frac{37 + 2 \times 42 + 3 \times 30 + \dots + 7}{150} = 2.11$$

b. Estimated standard error:

$$\sqrt{\frac{2.11}{150}} = 0.119$$

13.

$$E(X) = \sum x_i \cdot f(x_i; \theta)$$

$$E(\bar{x}) = \sum x_i \cdot f(x_i; \theta) = E(X)$$

$$\hat{\theta} = 3\bar{x}$$

$$E(\hat{\theta}) = 3E(\bar{x}) = E(3X) = \theta$$

So θ is an unbiased estimator

20.

a. When X is binomial rv,

$$\hat{p} = \frac{X}{n} = \frac{3}{20} = 0.15$$

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

so the estimator of part (a) is unbiased.

$$c. p = \left(1 - \frac{3}{20}\right)^5 = 0.44$$

21. a.

$$E(X^2) = V(X) + E(X)^2$$

$$= \beta^2 \Gamma(1 + 2/\alpha)$$

$$\bar{X}^2 = \beta^2 \Gamma(1 + 2/\alpha)$$

$$\beta = \frac{\bar{X}}{\Gamma(1 + 1/\alpha)}$$

$$s^2 = \left(\frac{\bar{X}}{\Gamma(1 + 1/\alpha)}\right)^2 \quad \Gamma(1 + 1/\alpha)^2 = \Gamma(1 + 2/\alpha)$$

$$\Gamma(1 + 1/\alpha)^2 / \Gamma(1 + 2/\alpha) = 1$$

$$b. [\Gamma(1.2)]^2 / \Gamma(1.4) = 0.95 \text{ so } \alpha = 5$$

$$b = \frac{28}{\Gamma(1.2)} = 28.0 / \Gamma(1.2)$$

21.

$$a. \tilde{E}(X) = \mu + \frac{1}{\lambda}$$

$$V(X) = \frac{1}{\lambda^2}$$

$$\bar{X} = \mu + \frac{1}{\lambda}$$

$$S^2 = \frac{1}{\lambda^2}$$

$$\lambda = \frac{1}{s}$$



$$\text{so } \hat{\theta} = \min(X_i), \hat{\lambda} = n / \sum (X_i - \min(X_i))$$

$$b. \hat{\theta} = \min(X_i) = 0.64$$

$$\hat{\lambda} = \frac{10}{\sum (X_i - \min(X_i))} = 0.202$$



$$32. a. F_Y(y) = P(Y \leq y)$$

$$= P(X_1 \leq y, X_2 \leq y, \dots, X_n \leq y)$$

$$= (P(X \leq y))^n$$

$$\text{For } (X_i \leq y) = \frac{y}{\theta}$$

$$F_Y(y) = \begin{cases} \left(\frac{y}{\theta}\right)^n & 0 < y < \theta \\ 0 & \text{otherwise} \end{cases}$$



$$\begin{aligned} b. \quad \tilde{E}(Y) &= \tilde{E}(Y) \\ &= \int_0^\theta y f_Y(y) dy \\ &= \frac{n}{n+1} \theta \end{aligned}$$

$$\begin{aligned} \tilde{E}\left(\frac{n+1}{n} Y\right) &= \frac{n+1}{n} \tilde{E}(Y) \\ &= \theta \end{aligned}$$

Thus, it is an unbiased estimator for θ
So the mle is biased but that $(n+1)\max(X_i)/n$
is unbiased.