

$$\frac{219.8}{27} = 8.1407$$

$$\bar{x} = 7.7$$

$$c. s = \sqrt{s^2} = \sqrt{\frac{1860.97 - \frac{(219.8)^2}{27}}{26}} = 1.660$$

$$d. \hat{\rho} = \frac{s}{\bar{x}} = \frac{1.660}{8.1407} = 0.2039$$

$$e. \frac{s}{\bar{x}} = \frac{1.660}{8.1407} = 0.2039$$

$$8. a. \hat{p} = \frac{20-12}{30} = 0.27$$

$$b. \hat{p}^2 = (0.27)^2 = 0.0729$$

$$9. a. n = 150$$

$$\hat{\mu} = \bar{x} = \frac{\sum x_i}{n} = \frac{(0)(18) + (1)(37) + (2)(42) + (3)(30) + (4)(12) + (5)(7) + (6)(2) + (7)(1)}{150}$$

$$= 2.11$$

$$b. s_{\bar{x}} = \frac{s}{\sqrt{n}} = \frac{\sqrt{11}}{\sqrt{150}} = 0.119$$

$$13. \mu = E(x) = \int_0^1 \frac{x}{4} (1 + \theta x) dx = \left[\frac{x^2}{8} + \frac{\theta x^3}{12} \right]_0^1 = \frac{1}{8} + \frac{\theta}{12}$$

$$\Rightarrow \hat{\theta} = 3\bar{x} \Rightarrow E(\hat{\theta}) = E(3\bar{x}) = 3E(\bar{x}) = 3\mu = 3\left(\frac{1}{8} + \frac{\theta}{12}\right) = \frac{3}{8} + \frac{\theta}{4} = \theta$$

$$b.2.20. a. p = \ln \left[\binom{n}{x} p^x (1-p)^{n-x} \right] = \ln \binom{n}{x} + x \ln p + (n-x) \ln (1-p)$$

$$p' = \frac{d}{dx} \left[\ln \binom{n}{x} + x \ln p + (n-x) \ln (1-p) \right] = \frac{x}{p} - \frac{n-x}{1-p} = 0$$

$$\Rightarrow x(1-p) = p(n-x)$$

$$\Rightarrow p = \frac{x}{n}$$

$$= \frac{3}{20}$$

$$= 0.15$$

b. It is an unbiased estimator of p .

$$c. (1-p)^5 = (1-\hat{p})^5 = (1-0.15)^5 = 0.4437$$

$$21. E(X) = B + (1 + \frac{1}{a}) \quad E(X^2) = V(X) + [E(X)]^2 = B^2 + (1 + \frac{2}{a})$$

$$\bar{x} = \hat{B} + (1 + \frac{1}{a}) \quad \frac{1}{n} \sum x_i^2 = \hat{B}^2 + (1 + \frac{2}{a})$$

$$\hat{B} = \frac{\bar{x}}{1 + \frac{1}{a}} \quad \bar{x}^2 = \hat{B}^2 + (1 + \frac{2}{a})$$

$$\frac{1}{n} \sum \frac{x_i^2}{\bar{x}^2} = \frac{1 + \frac{2}{a}}{1 + \frac{1}{a}}$$

$$\frac{1}{20} \left(\frac{16500}{282} \right) = 1.05 = \frac{1 + \frac{2}{a}}{1 + \frac{1}{a}} \quad \frac{1 + \frac{2}{a}}{1 + \frac{1}{a}} = 1.05$$

$$\frac{1}{a} = 0.2 \Rightarrow \hat{a} = 5 \quad \hat{B} = \frac{\bar{x}}{1 + \frac{1}{a}} = \frac{28}{1.2}$$



joint pdf (likelihood function) :

$$L(x_1, \dots, x_n; \lambda, \theta) = \begin{cases} \lambda^n e^{-\lambda \sum (x_i - \theta)} & x_1 \geq \theta, \dots, x_n \geq \theta \\ 0 & \text{otherwise} \end{cases}$$

$$\text{Likelihood} = \begin{cases} \lambda^n \exp(-\lambda \sum x_i) \exp(n\lambda\theta) & \min(x_i) \geq \theta \\ 0 & \min(x_i) < \theta \end{cases}$$

$$\hat{\theta} = \min(x_i)$$

$$\text{the log likelihood} = n \ln(\lambda) - \lambda \sum (x_i - \hat{\theta})$$

$$\hat{\lambda} = \frac{n}{\sum (x_i - \hat{\theta})} = \frac{n}{55.8 - 6 \cdot 4} = 0.202$$

b. $\theta = 0.64$ $\sum x_i = 55.8$ $\hat{\lambda} = \frac{10}{55.8 - 6 \cdot 4} = 0.202$

32. a. $F_Y(Y) = P(Y \leq y) = P(X_1 \leq y, \dots, X_n \leq y) = P(X_1 \leq y) \dots P(X_n \leq y)$
 $= \left(\frac{y}{\theta}\right)^n \text{ for } 0 \leq y \leq \theta$

$$f_Y(y) = \frac{ny^{n-1}}{\theta^n}$$

b. $E(Y) = \int_0^\theta y \cdot \frac{ny^{n-1}}{\theta^n} dy = \frac{n}{n+1} \theta$

$\hat{\theta} = Y$ is not unbiased, $\frac{n+1}{n} Y$ is since

$$E\left(\frac{n+1}{n} Y\right) = \frac{n+1}{n} E(Y) = \frac{n+1}{n} \cdot \frac{n}{n+1} \theta = \theta$$

