

Chapter 6

Section 6.1.

Ex. 1

(a) Use mean value: $\bar{x} = \frac{\sum x_i}{n} = \frac{219.8}{27} = 8.14$

(b) Use median value: $\hat{x} = 7.7$

(c) Use: $\sigma^2 = \frac{\sum (x_i - \bar{x})^2}{n-1} = \frac{\sum x_i^2 - 2\bar{x}\sum x_i + n\bar{x}^2}{n-1} = \frac{1860.94 - 2 \times 8.14 \times 219.8 + 27 \times 8.14^2}{27-1} = \frac{1860.94 - 3578.12 + 1811.18}{26} = \frac{1093.0}{26} = 42.04$

(d) Use: $\frac{4}{27} = 0.148$

(e) Use: $\frac{\sigma}{\bar{x}} = \frac{1.66}{8.14} = 0.204$

$\sigma = \sqrt{42.04} = 6.48$
 $\frac{\sigma}{\bar{x}} = \frac{6.48}{8.14} = 0.796$

Ex. (8)

(a) $1 - \frac{12}{8} = 0.85$

(b) It is clear that: $p = 0.85$

so: the answer is: $p^2 = 0.85^2 = 0.7225$

Ex. (9)

(a) $E(\bar{x}) = \mu = \bar{x} = \mu = E(x) = \frac{0 \times 8 + 1 \times 31 + 2 \times 42 + \dots + 7 \times 1}{150} = 2.11$

(b) ~~As for poisson distribution:~~

(b): $\sigma_x^2 = \mu$ so: $\sigma_{\bar{x}} = \frac{\sigma_x}{\sqrt{n}} = \frac{\sqrt{2.11}}{\sqrt{150}} = 0.119$

Ex. (13)

first calculate $\mu = E(x) = E(\bar{x}) = \int_{-1}^1 x f(x) dx = \int_{-1}^1 x \cdot (0.5(1+\theta x)) dx = \frac{\theta}{3}$

so: $\hat{\theta} = 3\bar{x}$, $E(\hat{\theta}) = E(3\bar{x}) = 3E(\bar{x}) = 3 \times \frac{\theta}{3} = \theta$

\Rightarrow unbiased estimator

Section 6.2

Ex. 20

(a) The example satisfy binomial distribution: so: $f(p) = p^3(1-p)^7$

$$\text{Then: } \ln f(p) = 3 \ln p + 7 \ln(1-p)$$

$$(\ln f(p))' = \frac{3}{p} - \frac{7}{1-p}$$

$$\text{let: } \frac{3}{p} - \frac{7}{1-p} = 0 \quad \text{then: } p = \frac{3}{10} = \frac{x}{n}$$

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

So: \hat{p} is unbiased

$$(c) (1-\hat{p})^5 = (0.85)^5 = 0.444$$

Ex. 21)

(a) We use moment method can derive the answer.

$$(b) E(X) = \beta \cdot \Gamma(1+1/\alpha)$$

$$E(X^2) = V(X) + E(X)^2 = \beta^2 (\Gamma(1+2/\alpha) - [\Gamma(1+1/\alpha)]^2) + \beta \cdot \Gamma(1+1/\alpha)$$

$$\text{Then: } \frac{\bar{X}^2}{\frac{1}{n} \sum X_i^2} = \frac{28^2}{825} = 0.95 = \frac{\Gamma(1+2/\alpha)}{\Gamma(1+1/\alpha)^2} = \frac{\Gamma(1+2/\alpha)}{\Gamma(1+1/\alpha)^2} \Rightarrow \alpha = 0.5$$

$$\text{then: } \beta = \frac{28}{\Gamma(1.2)}$$



Ex. 29.

(a) According to the problem: $\hat{\theta} = \min(X_i)$

$$f(x_1 x_2 \dots x_n) = \lambda e^{-\lambda(x_1 - \theta)} \cdot \lambda e^{-\lambda(x_2 - \theta)} \dots \lambda e^{-\lambda(x_n - \theta)}$$

$$= \lambda^n e^{-\lambda(\sum x_i - n\theta)}$$

differentiate $f(x_1 x_2 \dots x_n)$ w.r.t λ : $\ln f(x_1 \dots x_n) = n \ln \lambda - \lambda(\sum x_i - n\theta)$

$$(\ln f(x_1 x_2 \dots x_n))' = \frac{n}{\lambda} - \sum x_i + n\theta = 0$$

$$\text{so: } \lambda = \frac{n}{\sum x_i - n\theta} = \frac{n}{\sum x_i - n \cdot \min(x_i)}$$

(2) $\hat{\theta} = 0.64$

$$\hat{\lambda} = \frac{10}{\sum x_i - 10 \times 0.64} = 0.202$$

Ex. 32.

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

$$= \frac{y}{\theta} \cdot \frac{y}{\theta} \dots \frac{y}{\theta}$$

$$= \left(\frac{y}{\theta}\right)^n$$

$$\text{so: } f_Y(y) = \frac{d}{dy} \left(\frac{y}{\theta}\right)^n = \frac{n y^{n-1}}{\theta^n}$$

(b) $E(\hat{\theta}) = \int_0^{\theta} y f_Y(y) dy = \frac{n}{n+1} \max(x_i) \quad (\hat{\theta} = \max(x_i))$

so: $E\left(\frac{n+1}{n} \hat{\theta}\right) = \frac{n+1}{n} E(\hat{\theta}) = \theta$ is unbiased.

