

Sufficiency

$$1. f(p; x_1, \dots, x_n) = p^{x_1} (1-p)^{n-x_1} p^{x_2} (1-p)^{n-x_2} \dots p^{x_n} (1-p)^{n-x_n} \\ = p^{\sum x_i} (1-p)^{n^2 - \sum x_i}$$

$$\text{Let } h(x_1, \dots, x_n) = 1, \quad v(T, p) = p^T (1-p)^{n^2 - T}$$

$$\therefore f(p; x_1, \dots, x_n) = h(x_1, \dots, x_n) \cdot v(T, p)$$

$$\therefore T = \sum_{i=1}^n x_i \text{ is sufficient}$$

$$2. f(p; x_1, \dots, x_n) = \prod_{i=1}^n p (1-p)^{x_i} \\ = p^n (1-p)^{\sum x_i}$$

$$\text{Let } h(x_1, \dots, x_n) = 1, \quad v(T, p) = p^n (1-p)^T$$

$$\therefore f(p; x_1, \dots, x_n) = h(x_1, \dots, x_n) \cdot v(T, p)$$

$$T = \sum_{i=1}^n x_i \text{ is sufficient}$$

$$3. f(p; x_1, \dots, x_n) = \prod_{i=1}^n C_{x_i+r-1}^{r-1} p^r (1-p)^{x_i}$$

$$= \left(\prod_{i=1}^n C_{x_i+r-1}^{r-1} \right) p^{nr} (1-p)^{\sum x_i}$$

$$h(x_1, \dots, x_n)$$

$$\text{Let } \sum x_i = T, \quad v(T, p)$$

$$\therefore f(p; x_1, \dots, x_n) = h(x_1, \dots, x_n) \cdot v(T, p)$$

$$\therefore T = \sum_{i=1}^n x_i \text{ is sufficient}$$



$$\begin{aligned}
 4. f(\beta; x_1, \dots, x_n) &= \frac{\beta^\alpha}{\prod_{i=1}^n \Gamma(\alpha)} x_i^{\alpha-1} e^{-\beta x_i} I_{x_i > 0} \\
 &= \underbrace{\frac{\beta^{n\alpha}}{(\Gamma(\alpha))^n} \left(\prod_{i=1}^n x_i^{\alpha-1} \right)}_{h(x_1, \dots, x_n)} \underbrace{e^{-\beta \sum x_i}}_{V(T, \beta)} \quad T = \sum x_i
 \end{aligned}$$

$\therefore T = \sum_{i=1}^n x_i$ is sufficient

$$\begin{aligned}
 5. f(\beta; x_1, \dots, x_n) &= \underbrace{\frac{\beta^{n\alpha}}{(\Gamma(\alpha))^n} \left(\prod_{i=1}^n x_i^{\alpha-1} \right)}_{V(T, \alpha)} \underbrace{e^{-\beta \sum x_i}}_{h(x_1, \dots, x_n)} \quad T = \sum_{i=1}^n x_i
 \end{aligned}$$

$\therefore T = \sum_{i=1}^n x_i$ is sufficient

MLE

1. (a) $\hat{\beta} = \frac{5}{43}$

(b) $\hat{\beta} = \frac{3}{38}$

2. $X_1, \dots, X_n \sim U(0, \theta)$

$$L(\theta; x_1, \dots, x_n) = \prod_{i=1}^n \frac{1}{\theta} = \theta^{-n}$$

$$L(\theta; x_1, \dots, x_n) = -n \ln \theta$$

$$\frac{dL}{d\theta} = -\frac{n}{\theta}$$

$\therefore L(\theta; x_1, \dots, x_n) = \theta^{-n}$ is a decreasing function for $\theta \geq x_{(n)}$

$\therefore \hat{\theta} = x_{(n)}$ is a consistent sequence



$$3. X_1, \dots, X_n \sim N(\mu, \sigma^2)$$

$$L(\mu, \sigma; X_1, \dots, X_n) = \prod_{i=1}^n \left(\frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{(X_i - \mu)^2}{2\sigma^2}\right\} \right) = (2\pi\sigma^2)^{-\frac{n}{2}} \exp\left(-\frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \mu)^2\right)$$

$$l(\mu, \sigma; X_1, \dots, X_n) = -\frac{1}{2\sigma^2} \sum_{i=1}^n (X_i - \mu)^2 - \frac{n}{2} \ln \sigma^2 - \frac{n}{2} \ln 2\pi$$

$$\frac{dl}{d\mu} = \frac{1}{\sigma^2} \sum_{i=1}^n (X_i - \mu) = 0 \Rightarrow \hat{\mu} = \frac{\sum X_i}{n} = \bar{X}$$

$$\frac{dl}{d\sigma^2} = \frac{1}{2\sigma^4} \sum_{i=1}^n (X_i - \mu)^2 - \frac{n}{2\sigma^2} = 0 \Rightarrow \hat{\sigma}^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n}$$

$$\therefore 95\% \text{ MLE} = \hat{\mu} \pm 1.645 \hat{\sigma}$$

$$= \bar{X} \pm 1.645 \sqrt{\frac{\sum (X_i - \bar{X})^2}{n}}$$

$$4. \text{ MLE of } v = P(X > 2)$$

$$= P\left(Z > \frac{2 - \mu}{\sigma}\right) = 1 - \Phi\left(\frac{2 - \mu}{\sigma}\right)$$

$$5. f(x) = \frac{1}{\pi(1+(x-\theta)^2)}, \quad -\infty < x < \infty$$

$$L(\theta, X_1, \dots, X_n) = \frac{1}{\pi^n \prod_{i=1}^n (1+(X_i - \theta)^2)} = \frac{1}{\pi^n} \cdot \frac{1}{\left(\prod_{i=1}^n (1+(X_i - \theta)^2)\right)}$$

$$l(\theta, X_1, \dots, X_n) = -n \ln \pi - \sum_{i=1}^n \ln(1+(X_i - \theta)^2)$$

$$\frac{dl}{d\theta} = \sum_{i=1}^n \frac{2(X_i - \theta)}{1+(X_i - \theta)^2} = 0$$

$$\Rightarrow \frac{2(X + \theta)}{1+(X + \theta)^2} - \frac{2(X - \theta)}{1+(X - \theta)^2} = 0$$

$$(X + \theta) + (X + \theta)(X - \theta)^2 - (X - \theta) - (X - \theta)(X + \theta)^2 = 0$$

$$2\theta + (X + \theta)(X - \theta)[X - \theta - (X + \theta)] = 0$$

$$2\theta - 2\theta(X + \theta)(X - \theta) = 0$$

$$n = 20$$

$$2\theta(1 - X^2 + \theta^2) = 0$$

$$\bar{X} = -1.1465$$

$$2\theta(\theta^2 + (1 - X^2)) = 0$$

$$\therefore \hat{\theta} = \pm \sqrt{X^2 - 1}$$

$$\hat{\theta} \approx \pm 0.56$$



$$b. \sum_{i=1}^{20} x = 6(20) = 120 \quad F(x) = 1 - e^{-\lambda x}$$

$$p(x > 15) = e^{-\lambda 15}$$

$$L(\mu, x_{20}) = \frac{1}{\mu^{20}} \exp\left(-\frac{120}{\mu}\right) \exp\left(-\frac{15}{\mu}\right) \\ = \frac{1}{\mu^{20}} \exp\left(-\frac{135}{\mu}\right)$$

$$l(\mu, x_{20}) = 20 \ln \mu - 135/\mu$$

$$\frac{dl}{d\mu} = \frac{20}{\mu} - 135 = 0 \Rightarrow \hat{\mu} = 6.15$$

$$7. f(x) = \frac{e^{-\lambda} \lambda^x}{x!}$$

$$L(\lambda; x_1, \dots, x_n) = e^{-\lambda n} \lambda^{\sum x_i} \left(\prod_{i=1}^n \frac{1}{x_i!}\right)$$

$$l(\lambda; x_1, \dots, x_n) = -n \ln \lambda + \sum x_i \ln \lambda + \ln\left(\prod_{i=1}^n \frac{1}{x_i!}\right)$$

$$\frac{dl}{d\lambda} = -n + \frac{\sum x_i}{\lambda} = 0 \Rightarrow \hat{\lambda} = \frac{\sum x_i}{n} = \bar{x}$$

$$sd = \sqrt{\hat{\lambda}} = \sqrt{\bar{x}}$$

$$8. f(x) = \lambda e^{-\lambda x}$$

$$L(\lambda, x_1, \dots, x_n) = \lambda^n e^{-\lambda \sum x_i}$$

$$l(\lambda, x_1, \dots, x_n) = n \ln \lambda - \lambda \sum x_i$$

$$\frac{dl}{d\lambda} = \frac{n}{\lambda} - \sum x_i = 0 \Rightarrow \hat{\lambda} = \frac{n}{\sum x_i} = \frac{1}{\bar{x}}$$

$$MLE \text{ of median} = \left(\hat{\lambda} - \ln 2, \hat{\lambda} + \frac{1}{3} \right) = \left(\frac{1}{\bar{x}} - \ln 2, \frac{1}{\bar{x}} + \frac{1}{3} \right)$$

