

$$1. (a) X \sim M(n, k, p), L(p) = \binom{n-1}{k-1} p^k (1-p)^{n-k}$$

$$l(p) = \log \left[ \binom{n-1}{k-1} \right] + k \log(p) + \log(1-p)(n-k)$$

$$\therefore \frac{\Delta l}{\Delta p} = \frac{p}{1-p} - \frac{n-k}{1-p} = 0 \quad \therefore \hat{p} = \frac{p}{n} = \boxed{\frac{5}{43}}$$

$$(b) L(p, x_1, \dots, x_n) = p^{\sum_i^n} (1-p)^{n-\sum_i^n} x_i$$

$$\therefore l(p) = \sum_i^n x_i \log(p) + (n - \sum_i^n x_i) \log(1-p)$$

$$\therefore \frac{\Delta l}{\Delta p} = \frac{\sum_i^n x_i}{p} - \frac{n - \sum_i^n x_i}{1-p} = 0 \quad \therefore \hat{p} = \frac{\sum_i^n x_i}{n} = \boxed{\frac{3}{58}}$$

$$2. X_i \sim \text{Uniform}(0, \theta).$$

$$L(\theta, x_1, \dots, x_n) = \frac{1}{\theta^n} \Rightarrow l(\theta, x_1, \dots, x_n) = -n \ln(\theta)$$

$$\therefore \frac{\Delta l}{\Delta \theta} = -\frac{n}{\theta} = 0 \quad \therefore 0 \leq x_1 \leq \dots \leq x_n \leq \theta$$

$$l(\theta, x_1, \dots, x_n) = \frac{1}{\theta^n} \leq \frac{1}{x_1 \dots x_n} \quad \therefore \hat{\theta} = x_{(n)},$$

$$3. f(x) = \frac{1}{\sqrt{2\pi}\sigma} \cdot e^{-\frac{1}{2\sigma^2}(x-\mu)^2}$$

$$L(\mu, \sigma^2, x_1, \dots, x_n) = \left( \frac{1}{\sqrt{2\pi}\sigma} \right)^n \cdot e^{-\frac{1}{2\sigma^2}[(x_1-\mu)^2 + \dots + (x_n-\mu)^2]}$$

$$l(\mu, \sigma^2, x_1, \dots, x_n) = n \log \left( \frac{1}{\sqrt{2\pi}\sigma} \right) - \frac{n}{2} \log(\sigma^2) - \frac{1}{2\sigma^2} \sum_i^n (x_i - \mu)^2$$

$$\frac{\Delta l}{\Delta \mu} = \frac{1}{\sigma^2} \sum_i^n (x_i - \mu) = 0, \quad \frac{\Delta l}{\Delta \sigma^2} = \frac{1}{2\sigma^4} \sum_i^n (x_i - \mu)^2 = 0.$$

$$\therefore \hat{\mu} = \frac{1}{n} \sum_{i=1}^n x_i \quad \hat{\sigma}^2 = \frac{n}{2} \overline{(x_i - \mu)^2}$$

$$\therefore \text{MLE of 95\% quantile. } \hat{\mu} \pm 1.96 \hat{\sigma} = \frac{1}{n} \sum_{i=1}^n x_i + 1.96 \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2}$$

$$4. V = P(X > z) = P(Z > \frac{z-\mu}{\sigma}) = 1 - P(Z \leq \frac{z-\mu}{\sigma}) = 1 - \Phi(\frac{z-\mu}{\sigma})$$

$$\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \mu)^2 / n \Rightarrow \hat{\sigma} = \sqrt{\frac{1}{n} \sum_{i=1}^n (X_i - \mu)^2 / n}$$

$$5. L(\theta, x) = \frac{1}{\pi^n} \cdot \frac{1}{[1+(x_1-\theta)^2] \cdots [1+(x_n-\theta)^2]}, \quad l(\theta, x) = -n \log \pi$$

$$\frac{\partial l}{\partial \theta} = \sum_{i=1}^n \frac{2(x_i-\theta)}{1+(x_i-\theta)^2} = 0.$$

MLE of  $\theta$  can be derived by given inputs of  $X$  by  $\frac{\partial l}{\partial \theta} = 0$

$$6. \because f(x) = \lambda e^{-\lambda x} \quad L(\lambda, x_1, \dots, x_n) = \lambda^n e^{-\lambda \sum_{i=1}^n x_i} \quad l(\lambda, x_1, \dots, x_n)$$

$$= n \log(\lambda) + \sum_{i=1}^n x_i \quad \frac{\partial l}{\partial \lambda} = \frac{n}{\lambda} - \sum_{i=1}^n x_i = 0$$

$$\therefore \lambda = \frac{n}{\sum_{i=1}^n x_i} = \frac{21}{15+6+20} = \frac{21}{45} = \frac{7}{15} \quad \lambda = \frac{1}{\lambda} = \frac{15}{7}.$$

$$7. \because f(x) = \frac{\lambda^{x_i}}{x_i!} e^{-\lambda}, \quad V(x) = E(x) = \lambda,$$

$$\Rightarrow L(\lambda, x_1, \dots, x_n) = \lambda^{\sum_{i=1}^n x_i} \cdot [x_1! \cdots x_n!]^{-1} e^{-n\lambda}$$

$$\Rightarrow l(\lambda, x_1, \dots, x_n) = \sum_{i=1}^n x_i \log(\lambda) - \log(x_1! \cdots x_n!) - n\lambda.$$

$$\frac{\partial l}{\partial \lambda} = \sum_{i=1}^n \frac{x_i}{\lambda} - n = 0 \Rightarrow \lambda = \frac{\sum_{i=1}^n x_i}{n}.$$

$$8. \text{ Assume } \alpha \text{ median} \quad \therefore f(x) = \lambda e^{-\lambda x}$$

$$\therefore \int_0^\infty \lambda e^{-\lambda x} dx = e^{-\frac{\alpha}{\lambda}} + 1 = 0.5 \Rightarrow e^{-\frac{\alpha}{\lambda}} = 0.5 \Rightarrow \alpha = \ln 2 \lambda.$$

$$\therefore \hat{\lambda} = \frac{n}{\sum_{i=1}^n x_i} \quad \therefore \hat{\alpha} = \ln 2 \cdot \frac{n}{\sum_{i=1}^n x_i}.$$

$$1. T = \sum_{i=1}^n x_i \quad \because f(x, p) = p^x = p^x (1-p)^{1-x}$$

$$\begin{aligned} L(p, x) &= p^{\sum x_i} (1-p)^{n-\sum x_i} = p^T (1-p)^{n-T} \quad \left\{ M(x)=1 \right. \\ &\quad \left. V(t, p) = p^{t(x)} (1-p)^{n-t(x)} \right. \end{aligned}$$

$$2. \because f(x, p) = p (1-p)^{x-1}$$

$$L(p, x) = p^n (1-p)^{\sum x_i - n} = p^n (1-p)^{\sum x_i - n} \quad \left\{ M(x)=1 \right. \\ \left. V(t, p) = p^n (1-p)^{-n+t(x)} \right.$$

$$3. \because f(x, r, p) = \binom{x+r-1}{x-1} p^x (1-p)^r$$

$$L(r, p, x) = p^{\sum x_i} (1-p)^{nr} \left[ \binom{x_1+r-1}{x_1-1} \cdots \binom{x_n+r-1}{x_n-1} \right] \quad \left\{ M(x)=1 \right. \\ \left. V(t, r, p) = p^{t(x)} (1-p)^{nr} \right.$$

$$4. \because f(x, \alpha, \beta) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}$$

$$L(x, \alpha, \beta) = \left( \frac{\beta^\alpha}{\Gamma(\alpha)} \right)^n e^{-\beta \sum x_i} (x_1 \cdots x_n)^{\alpha-1} \quad \left\{ M(x) = \beta^{-n} \left( \prod_{i=1}^n x_i \right)^{\alpha-1} \right. \\ \left. V(t, \beta) = \beta^{\alpha t} e^{-\beta t(x)} \right.$$

$$5. \left\{ \begin{array}{l} M(x) = e^{-\beta \sum x_i} \\ V(t, \alpha) = (t)^{\alpha-1} \left( \frac{\beta^\alpha}{\Gamma(\alpha)} \right)^n \end{array} \right.$$