

1) Let  $X_1, X_2, \dots, X_n$  be a random sample from distribution

$$\frac{x_i}{P} \begin{array}{c|cc} 0 & 1 \\ 1-p & p \end{array} \text{ using the Method of Moments find estimator of } p.$$

After a series of  $n = 10$  trials this observations where made

0, 1, 1, 0, 1, 1, 1, 0, 0, 1 find the estimate of  $p$ .

$$\begin{cases} M_1 = E(X) = \bar{X} = p \\ M_2 = E(X^2) = \frac{1}{n} \sum X_i^2 = p \end{cases}$$

$$M_1 - M_2 \implies 0 = \bar{X} - \frac{1}{n} \sum X_i^2$$

$$\hat{p}_{MM} = \bar{X} = \frac{1}{n} \sum X_i^2$$

$$L(p) = p^{\sum X_i} \cdot (1-p)^{n-\sum X_i}$$

$$\hat{p}_{MLE} = \frac{1}{n} \sum X_i = \bar{X}$$

$$\therefore \{x_1, \dots, x_n\} = \{0, 1, 1, 0, 1, 1, 1, 0, 0, 1\}$$

$$\therefore \hat{p}_{MLE} = \frac{6}{10}, \hat{p}_{MM} = \frac{6}{10}$$

2) Let  $X_1, X_2, \dots, X_n$  be a random sample from distribution

$$\frac{x_i}{P} \begin{array}{c|ccc} 0 & 1 & 3 \\ 1-p_1-p_2 & p_1 & p_2 \end{array} \text{ using the Method of Moments find estimators of } p_1 \text{ and } p_2$$

After a series of  $n = 10$  trials this observations where made 0, 1, 3, 0, 1, 1, 1, 3, 3, 1

find the estimates of  $p_1$  and  $p_2$ . Are the obtained estimators unbiased?

$$\begin{cases} M_1 = E(X) = \bar{X} = p_1 + 3p_2 \\ M_2 = E(X^2) = \frac{1}{n} \sum X_i^2 = p_1 + 9p_2 \end{cases} \begin{cases} M_2 - M_1 \implies p_2 = \frac{1}{6} (\frac{1}{n} \sum X_i^2 - \bar{X}) \\ 3M_1 - M_2 \implies p_1 = \frac{1}{2} (3\bar{X} - \frac{1}{n} \sum X_i^2) \end{cases}$$

$$\therefore \{x_1, \dots, x_n\} = \{0, 1, 3, 0, 1, 1, 1, 3, 3, 1\}, \bar{X} = \frac{14}{10}, \frac{1}{n} \sum X_i^2 = \frac{32}{10}$$

$$\therefore \hat{p}_1 = \frac{1}{2}, \hat{p}_2 = \frac{3}{10}$$

$$S^2 = \frac{1}{n-1} \cdot \sum (X_i - \bar{X})^2 = \frac{62}{45}$$

$$E(X) = \frac{1}{2} + \frac{9}{10} = \frac{14}{10} = \bar{X}$$

$$D(x) = \frac{1}{2} + \frac{27}{10} = \frac{16}{5} > S^2$$

so,  $\hat{\theta}$  is better

3) Consider a random sample  $X_1, X_2, \dots, X_n$  from the shifted exponential pdf

$$f(x) \begin{cases} \lambda e^{-\lambda(x-\theta)} & x \geq \theta \\ 0 & x < \theta \end{cases}$$

a. Obtain the maximum likelihood estimators of  $\theta$  and  $\lambda$ .

b. Obtain the method of moment estimators of  $\theta$  and  $\lambda$ .

c. If  $n = 10$  time observations are made,

resulting in the values 3.11, 0.64, 2.55, 2.20, 5.44, 3.42, 10.39, 8.93, 17.82, and 1.30,

calculate the estimates of  $\theta$  and  $\lambda$ .

$$a) L(\theta, \lambda) = \prod_{i=1}^n \lambda e^{-\lambda(x_i-\theta)} = \lambda^n \prod_{i=1}^n e^{-\lambda(x_i-\theta)}$$

$$\ln L = n \ln \lambda + (-\lambda) \sum_{i=1}^n (x_i - \theta) = n \ln \lambda + \lambda n \theta - \lambda \sum_{i=1}^n (x_i)$$

$$\begin{cases} \frac{\partial \ln L}{\partial \theta} = \lambda n > 0 & \therefore \theta \uparrow, L(\theta) \uparrow \therefore \theta = X_{\min} \implies \hat{\theta}_{MLE} = \{X_1, \dots, X_n\}_{\min} \\ \frac{\partial \ln L}{\partial \lambda} = \frac{n}{\lambda} + n\theta - \sum_{i=1}^n x_i = 0 & \therefore \lambda = \frac{n}{\sum_{i=1}^n x_i - n\theta} = \frac{1}{\bar{x} - \theta} \implies \hat{\lambda}_{MLE} = \frac{1}{\bar{x} - \hat{\theta}_{MLE}} \end{cases}$$

$$b) \mu_1 = E(x) = \int_{-\infty}^{+\infty} x f(x) dx = \int_{\theta}^{+\infty} x \lambda e^{-\lambda(x-\theta)} dx = -x e^{-\lambda(x-\theta)} \Big|_{\theta}^{+\infty} - \int_{\theta}^{+\infty} \lambda e^{-\lambda(x-\theta)} dx$$

$$= \lim_{x \rightarrow +\infty} x e^{-\lambda(x-\theta)} - (-\theta) + \left( -\frac{1}{\lambda} e^{-\lambda(x-\theta)} \Big|_{\theta}^{+\infty} \right) = 0 + \theta + \left( \lim_{x \rightarrow +\infty} e^{-\lambda(x-\theta)} - \frac{1}{\lambda} \right)$$

$$= \theta + \frac{1}{\lambda}$$

$$\mu_2 = E(x^2) = \int_{-\infty}^{+\infty} x^2 f(x) dx = \int_{\theta}^{+\infty} x^2 \lambda e^{-\lambda(x-\theta)} dx$$

$$= -x^2 e^{-\lambda(x-\theta)} \Big|_{\theta}^{+\infty} - \int_{\theta}^{+\infty} -e^{-\lambda(x-\theta)} d(x^2)$$

$$= \lim_{x \rightarrow +\infty} x^2 e^{-\lambda(x-\theta)} - (-\theta^2) + 2 \int_{\theta}^{+\infty} x \lambda e^{-\lambda(x-\theta)} dx$$

$$= 0 + \theta^2 + 2\left(\theta + \frac{1}{\lambda}\right) = \theta^2 + 2\theta + \frac{2}{\lambda}$$

$$A_1 = \bar{x} = \mu_1 \implies \hat{\theta}_{MM} = \bar{x} - \frac{1}{\lambda}$$

$$A_2 = \frac{1}{n} \sum x_i^2 = \mu_2 \implies \hat{\lambda}_{MM} = \frac{2}{\frac{1}{n} \sum x_i^2 - \theta^2 - 2\theta}$$

$$c) \hat{\theta}_{MLE} = \min(X) = 0.64, \hat{\lambda}_{MLE} = \frac{1}{\bar{x} - \hat{\theta}_{MLE}} = \frac{1}{5.58 - 0.64} \approx 0.2$$

$$\hat{\theta}_{MM} = \infty, \hat{\lambda}_{MM} = 0$$