

University of Technology Sydney
School of Mathematical and Physical Sciences

Mathematical Statistics (37262) –
 Class 6 Preparation Work
 SOLUTIONS

1. i) We know that, for $Y \sim \text{Poi}(\lambda)$, $E(Y) = \lambda$, hence the first moment of the distribution is $m_1 = \lambda$.

The first sample moment is $s_1 = \bar{y} = \frac{y_1 + y_2 + \dots + y_n}{n}$.

Applying the method of moments and matching the first moment, we obtain $m_1 = s_1$ and hence $\hat{\lambda}_{MM} = \frac{y_1 + y_2 + \dots + y_n}{n}$.

$$\text{ii) } L(\{y_1, y_2, \dots, y_n\} | \lambda) = \prod_{i=1}^n \left(\frac{e^{-\lambda} \lambda^{y_i}}{y_i!} \right) = \frac{\prod_{i=1}^n (e^{-\lambda} \lambda^{y_i})}{\prod_{i=1}^n y_i!} = \frac{e^{-n\lambda} \lambda^{\sum_{i=1}^n y_i}}{\prod_{i=1}^n y_i!}.$$

$$\begin{aligned} \text{iii) } \ell(\{y_1, y_2, \dots, y_n\} | \lambda) &= \ln(L(\{y_1, y_2, \dots, y_n\} | \lambda)) \\ &= -n\lambda + \ln(\lambda) \sum_{i=1}^n y_i - \ln \left(\prod_{i=1}^n y_i! \right). \end{aligned}$$

- iv) We maximise the loglikelihood (and hence also the likelihood) but finding when the derivative with respect to λ is equal to zero.

$$\frac{\partial}{\partial \lambda} \ell(\{y_1, y_2, \dots, y_n\} | \lambda) = -n + \frac{1}{\lambda} \sum_{i=1}^n y_i.$$

Solving $-n + \frac{1}{\lambda} \sum_{i=1}^n y_i = 0$ gives $\hat{\lambda}_{MLE} = \frac{1}{n} \sum_{i=1}^n y_i$ and hence $\hat{\lambda}_{MLE} = \hat{\lambda}_{MM}$.

2.

$$\text{i)} \quad s_1 = \bar{z} = \frac{1+4+3+3+3}{5} = \frac{14}{5} = 2.8$$

and

$$s_2 = \frac{1^2 + 4^2 + 3^2 + 3^2 + 3^2}{5} = \frac{44}{5} = 8.8.$$

ii) For $Z \sim \text{Bin}(n, p)$, we have

$$E(Z) = np \text{ and } E(Z^2) = \text{Var}(Z) + E(Z)^2 = np(1-p) + (np)^2$$

Applying the method of moments and matching the first two moments, we obtain $np = 2.8$ and $np(1-p) + (np)^2 = 8.8$.

Solving these gives $2.8(1-p) + 2.8^2 = 8.8$ hence $\frac{8.8 - 2.8^2}{2.8} = 1-p$, so

$$\frac{8.8 - 2.8^2}{2.8} = 1-p.$$

This gives $\hat{p}_{MM} = 1 - \frac{8.8}{2.8} + 2.8 = \frac{23}{35}$ and hence $\hat{n}_{MM} = 2.8 \left(\frac{35}{23} \right) = \frac{98}{23}$.

$$\text{iii)} \quad L(\{z_1, z_2, \dots, z_5\} | p) = \prod_{i=1}^5 \left(\frac{5!}{z_i! (5-z_i)!} p^{z_i} (1-p)^{(5-z_i)} \right) \text{ hence}$$

$$L(\{z_1, z_2, \dots, z_5\} | p) = \prod_{i=1}^5 \left(\frac{5!}{z_i! (5-z_i)!} \right) p^{\sum_{i=1}^5 z_i} (1-p)^{\sum_{i=1}^5 (5-z_i)}.$$

This gives

$$\ell(\{z_1, z_2, \dots, z_5\} | p) = \sum_{i=1}^5 z_i \ln(p) + \sum_{i=1}^5 (5-z_i) \ln(1-p) + \ln \left(\prod_{i=1}^5 \left(\frac{5!}{z_i! (5-z_i)!} \right) \right)$$

. Maximising this, we solve $\frac{\partial}{\partial p} \ell(\{z_1, z_2, \dots, z_5\} | p) = 0$.

$$\frac{\partial}{\partial p} \ell(\{z_1, z_2, \dots, z_5\} | p) = \frac{1}{p} \sum_{i=1}^5 z_i - \frac{1}{1-p} \sum_{i=1}^5 (5-z_i) \text{ hence we solve}$$

$$(1-p) \sum_{i=1}^5 z_i - p \sum_{i=1}^5 (5-z_i) = 0, \text{ which gives}$$

$$\sum_{i=1}^5 z_i - p \sum_{i=1}^5 z_i - p \sum_{i=1}^5 (5-z_i) = \sum_{i=1}^5 z_i - p \sum_{i=1}^5 z_i - p \left(25 - \sum_{i=1}^5 z_i \right) = 0.$$

$$\sum_{i=1}^5 z_i - 25p = 0 \text{ hence } \hat{p}_{MLE} = \frac{1}{25} \sum_{i=1}^5 z_i = \frac{14}{25} = 0.56.$$

iv)

$$L(\{z_1, z_2, \dots, z_5\} | n) = \prod_{i=1}^5 \left(\frac{n!}{z_i! (n-z_i)!} 0.5^{z_i} (0.5)^{(n-z_i)} \right) = \prod_{i=1}^5 \left(\frac{n!}{z_i! (n-z_i)!} 0.5^n \right)$$

This cannot readily be maximised by a derivative since the factorial function is only defined for integer values hence is differentiable nowhere.

We have seen an average of $\frac{14}{5} = 2.8$ on the binomial trials. Given that

we now know that $p = 0.5$, this implies that n should be close to 5.6.

We therefore compare $n=5$ and $n=6$ and select whichever gives a larger likelihood.

$$L(\{z_1, z_2, \dots, z_5\} | n=5) = \prod_{i=1}^5 \left(\frac{5!}{z_i! (5-z_i)!} 0.5^5 \right) \approx 0.000745$$

$$L(\{z_1, z_2, \dots, z_5\} | n=6) = \prod_{i=1}^5 \left(\frac{6!}{z_i! (6-z_i)!} 0.5^6 \right) \approx 0.000671$$

hence we conclude that $\hat{n}_{MLE} = 5$.