

6.4

$$\begin{aligned} E(\hat{\theta}_1) &= E\left(\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n}\right) = \frac{1}{n} E\left(\sum_{i=1}^n x_i^2 - n\bar{x}^2\right) \\ &= \frac{1}{n} (n\sigma^2 + n\mu^2 - \sigma^2 - n\mu^2) = \frac{n-1}{n} \sigma^2 \end{aligned}$$

$$\begin{aligned} E(\hat{\theta}_2) &= E\left(\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}\right) = \frac{1}{n-1} E\left(\sum_{i=1}^n x_i^2 - n\bar{x}^2\right) \\ &= \frac{1}{n-1} (n\sigma^2 + n\mu^2 - \sigma^2 - n\mu^2) = \sigma^2 \end{aligned}$$

\Rightarrow 故： $\hat{\theta}_2 = \sum_{i=1}^n (x_i - \bar{x})^2 / (n-1)$ 為 σ^2 之 不偏估計量
 $\hat{\theta}_1 = \sum_{i=1}^n (x_i - \bar{x})^2 / n$ 為 σ^2 之 有偏估計量