

Ch6 例 6-4

$$E(X_i) = \mu, V(X) = \sigma^2 = E(X_i^2) - \mu^2$$

$$E(\bar{X}) = \mu, V(\bar{X}) = \frac{\sigma^2}{n} = E(\bar{X}^2) - \mu^2$$

$$E(\hat{\theta}_1) = E\left(\frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n}\right) = \frac{1}{n} E\left(\sum_{i=1}^n X_i^2 - n\bar{X}^2\right)$$

$$= \frac{1}{n} (n\sigma^2 + \cancel{n\mu^2} - \sigma^2 - \cancel{n\mu^2}) = \frac{n-1}{n} \sigma^2$$

$$E(\hat{\theta}_2) = E\left(\frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1}\right) = \frac{1}{n-1} E\left(\sum_{i=1}^n X_i^2 - n\bar{X}^2\right)$$

$$= \frac{1}{n-1} (n\sigma^2 + \cancel{n\mu^2} - \sigma^2 - \cancel{n\mu^2}) = \sigma^2$$

由上推導可知 $\hat{\theta}_2 = \sum_{i=1}^n (X_i - \bar{X})^2 / (n-1)$ 為母體 $V(X)$ 之 unbiased 估計量，而 $\hat{\theta}_1 = \sum_{i=1}^n (X_i - \bar{X})^2 / n$ 為母體 $V(X)$ 之 biased 估計量