

Example 5.1 $X_i \sim N(\mu, \sigma^2)$ i.i.d.

mse of S^2 $\text{mse}(S^2; \sigma^2) = (E[S^2] - \sigma^2)^2 + \text{Var}(S^2)$

↑ estimator ↑ true value of population parameter

\uparrow $= 0^2 + \frac{2\sigma^4}{n-1} = \frac{2\sigma^4}{n-1}$
 ch. 4 (4.3.1)

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

$$\tilde{S}^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2 = \frac{n-1}{n} S^2$$

$\text{mse}(\tilde{S}^2; \sigma^2) = ?$

$$E \tilde{S}^2 = E\left[\frac{n-1}{n} S^2\right] = \frac{n-1}{n} E S^2 = \frac{n-1}{n} \sigma^2$$

$$\text{Var} \tilde{S}^2 = \text{Var}\left(\frac{n-1}{n} S^2\right) = \left(\frac{n-1}{n}\right)^2 \text{Var} S^2 = \left(\frac{n-1}{n}\right)^2 \cdot \frac{2\sigma^4}{n-1}$$

$$\text{mse}(\tilde{S}^2; \sigma^2) = (E \tilde{S}^2 - \sigma^2)^2 + \text{Var} \tilde{S}^2$$

$$= \left(-\frac{1}{n} \sigma^2\right)^2 + \frac{2(n-1)}{n^2} \sigma^4 = \frac{2n-1}{n^2} \sigma^4 < \frac{2\sigma^4}{n-1}$$

So $\text{mse}(\tilde{S}^2; \sigma^2) < \text{mse}(S^2; \sigma^2)$.

Example 5.3 $X_i \sim N(\mu, \sigma^2)$

$\theta = \sigma^2$ is the parameter of interest.

(because $\sum x_i^2$ estimates σ^2 , not σ .)

$$I(\theta) = n \cdot i(\theta) = n \cdot \mathbb{E}_X \left[-\frac{\partial^2}{\partial \theta^2} \log f_X(x; \theta) \right]$$

$$f_X(x; \theta) = \frac{1}{\sqrt{2\pi\theta}} \exp\left(-\frac{1}{2\theta} (x-\mu)^2\right)$$

$$\log f_X(x; \theta) = -\frac{1}{2} \log \theta - \frac{1}{2\theta} (x-\mu)^2 + \text{const.}$$

$\log x \propto$
" "
 $\propto \log x$

$$\frac{\partial}{\partial \theta} \log f_X(x; \theta) = -\frac{1}{2\theta} + \frac{1}{2\theta^2} (x-\mu)^2$$

$$\frac{\partial^2}{\partial \theta^2} \log f_X(x; \theta) = \frac{1}{2\theta^2} - \frac{2}{2\theta^3} (x-\mu)^2$$

$$\mathbb{E} I(\theta) = n \cdot i(\theta) = n \cdot \mathbb{E} \left[-\left(\frac{1}{2\theta^2} - \frac{1}{\theta^3} (x-\mu)^2 \right) \right]$$

$$= -\frac{n}{2\theta^2} + \frac{n}{\theta^3} \underbrace{\mathbb{E}[(x-\mu)^2]}_{=\theta} = \frac{n}{2\theta^2}$$

So, the Cramér-Rao lower bound says

$$\text{Var}(T) \geq \frac{1}{I(\theta)} = \frac{2\theta^2}{n} = \frac{2\sigma^4}{n}$$

Note: We know $\text{Var}\left(\sum_{i=1}^n x_i^2\right) = \frac{2\sigma^4}{n}$, so $\sum x_i^2$ achieves the Cramér-Rao lower bound.