

例 1.6.4 正態分布 $N(\mu, \sigma^2)$ の母平均 μ の推定

$$\text{依り } E(X_i) = \mu \quad V(X_i) = \sigma^2 = E(X_i^2) - \mu^2$$

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

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

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

$$= \sigma^2 + \mu^2 - \frac{\sigma^2}{n} - \mu^2$$

$$= \sigma^2 - \frac{\sigma^2}{n}$$

$$= \frac{n-1}{n} \sigma^2 \Rightarrow E(\hat{\sigma}^2) \text{ は } \sigma^2 \text{ の有偏推定量}$$

\Rightarrow 偏推定量

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

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

$$= \frac{n\sigma^2}{n-1} - \frac{\sigma^2}{n-1}$$

$$= \sigma^2 \Rightarrow E(\hat{\sigma}^2) \text{ は } \sigma^2 \text{ の無偏推定量}$$

\Rightarrow 無偏推定量

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