

< 復元抽出による標本調査 2 >

前のページの定理 14 を証明する。

< (1) の証明 >

$$E[X_k] = \sum_{i=1}^N y_i P(X_k = y_i) = \frac{1}{N} \sum_{i=1}^N y_i = \mu \quad (\text{母平均})$$

$$E[(X_k - \mu)^2] = \sum_{i=1}^N (y_i - \mu)^2 P(X_k = y_i) = \frac{1}{N} \sum_{i=1}^N (y_i - \mu)^2 = \sigma^2 \quad (\text{母分散})$$

< (2) の証明 >

$$E[\bar{X}] = E\left[\frac{1}{n} \sum_{k=1}^n X_k\right] = \frac{1}{n} \sum_{k=1}^n E[X_k] = \frac{1}{n} \sum_{k=1}^n \mu = \mu$$

$$\begin{aligned} E\left[\sum_{k=1}^n (X_k - \bar{X})^2\right] &= \sum_{k=1}^n E[(X_k - \mu + \mu - \bar{X})^2] \\ &= \sum_{k=1}^n \{E[(X_k - \mu)^2] + 2E[(X_k - \mu)(\mu - \bar{X})] + E[(\mu - \bar{X})^2]\} \end{aligned}$$

ここで

$$\begin{aligned} E[(X_k - \mu)(\mu - \bar{X})] &= E\left[(X_k - \mu) \left\{\frac{1}{n} \sum_{i=1}^n (\mu - X_i)\right\}\right] \\ \frac{1}{n} \sum_{i=1}^n E[(X_k - \mu)(\mu - X_i)] &= \frac{1}{n} \left\{-E[(X_k - \mu)^2] + \sum_{i \neq k} E[(X_k - \mu)]E[(\mu - X_i)]\right\} \\ &= -\frac{1}{n} E[(X_k - \mu)^2] = -\frac{\sigma^2}{n} \end{aligned}$$

また

$$\begin{aligned} E[(\mu - \bar{X})^2] &= E\left[\left\{\frac{1}{n} \sum_{i=1}^n (\mu - X_i)\right\}^2\right] \\ &= \frac{1}{n^2} \left\{\sum_{i=1}^n E[(\mu - X_i)^2] + \sum_{l \neq i} E[(\mu - X_i)(\mu - X_l)]\right\} = \frac{1}{n^2} \left(\sum_{i=1}^n \sigma^2\right) = \frac{\sigma^2}{n} \end{aligned}$$

となる。よって

$$\begin{aligned} E\left[\sum_{k=1}^n (X_k - \bar{X})^2\right] &= \sum_{k=1}^n \left\{\sigma^2 + 2\left(-\frac{\sigma^2}{n}\right) + \frac{\sigma^2}{n}\right\} = n\sigma^2 \left(1 - \frac{1}{n}\right) \\ &= (n-1)\sigma^2 \end{aligned}$$

(証明終)