

Ch.6 T21

$X \sim N(\mu_1, \sigma^2)$ ,  $Y \sim N(\mu_2, \sigma^2)$  且互相独立

$$\alpha(\bar{X} - \mu_1) \sim N(0, \frac{\alpha^2 \sigma^2}{m}), \quad \beta(\bar{Y} - \mu_2) \sim N(0, \frac{\beta^2 \sigma^2}{n})$$

独立 + 正态可加  $\Rightarrow \alpha(\bar{X} - \mu_1) + \beta(\bar{Y} - \mu_2) \sim N(0, (\frac{\alpha^2}{m} + \frac{\beta^2}{n})\sigma^2)$

$$\frac{(m-1)S_{1m}^2}{\sigma^2} \sim \chi^2(m-1), \quad \frac{(n-1)S_{2n}^2}{\sigma^2} \sim \chi^2(n-1)$$

由于  $S_{1m}^2$  和  $S_{2n}^2$  独立, 故  $\frac{(m-1)S_{1m}^2 + (n-1)S_{2n}^2}{\sigma^2} \sim \chi^2(m+n-2)$

$$\text{令 } W = \frac{\alpha(\bar{X} - \mu_1) + \beta(\bar{Y} - \mu_2)}{\sqrt{\frac{\alpha^2}{m} + \frac{\beta^2}{n}} \sigma}, \quad \text{则 } W \sim N(0, 1).$$

$$\text{故 } T = \frac{W}{\sqrt{\frac{(m-1)S_{1m}^2 + (n-1)S_{2n}^2}{m+n-2} \cdot \frac{1}{\sigma}}} = \frac{\alpha(\bar{X} - \mu_1) + \beta(\bar{Y} - \mu_2)}{\sqrt{\frac{(m-1)S_{1m}^2 + (n-1)S_{2n}^2}{m+n-2} \cdot \sqrt{\frac{\alpha^2}{m} + \frac{\beta^2}{n}}}} \sim t(m+n-2)$$

Ch.7 T6

$$\begin{aligned} (1) \quad EX &= \int_0^{\theta} x \cdot \frac{1}{2\theta} dx + \int_{\theta}^1 x \cdot \frac{1}{2(1-\theta)} dx \\ &= \frac{1}{2\theta} \left( \frac{1}{2}\theta^2 - 0 \right) + \frac{1}{2(1-\theta)} \left( \frac{1}{2} - \frac{1}{2}\theta^2 \right) \\ &= \frac{\theta}{4} + \frac{1+\theta}{4} = \frac{1+2\theta}{4} \end{aligned}$$

$\bar{X}$  是  $EX$  矩估计, 故  $\bar{X} = \frac{1+2\hat{\theta}}{4}$

$$\text{即 } \hat{\theta} = \frac{4\bar{X} - 1}{2}.$$

$$(2) \quad E4\bar{X}^2 = 4E\bar{X}^2$$

$$\begin{aligned} EX^2 &= \int_0^{\theta} x^2 \cdot \frac{1}{2\theta} dx + \int_{\theta}^1 x^2 \cdot \frac{1}{2(1-\theta)} dx \\ &= \frac{2\theta^2 + \theta + 1}{6} \end{aligned}$$

$$E\bar{X}^2 = (E\bar{X})^2 + D(\bar{X})$$

$$= (EX)^2 + \frac{1}{n} DX$$

$$= \frac{4\theta^2 + \theta + 1}{16} + \frac{1}{n} (EX^2 - (EX)^2)$$

$$= \frac{\theta^2}{4} + \frac{\theta}{4} + \frac{1}{16} + \frac{1}{n} \left( \frac{\theta^2}{3} + \frac{\theta}{6} + \frac{1}{6} - \frac{1}{16} - \frac{\theta}{4} - \frac{\theta^2}{4} \right)$$

$4E\bar{X}^2$  显然不等于  $\theta^2 \dots$  故不是无偏估计

T13. 因为是随机的, 可以看成独立同分布.

$$L(\theta) = (\theta^2)^{n_1} (2\theta(1-\theta))^{n_2} ((1-\theta)^2)^{n_3}$$

$$l = \ln L(\theta) = 2n_1 \ln \theta + n_2 \ln 2\theta(1-\theta) + 2n_3 \ln(1-\theta)$$

$$= (2n_1 + n_2) \ln \theta + (2n_3 + n_2) \ln(1-\theta) + n_2 \ln 2$$

$$l'_{\theta} = \frac{2n_1 + n_2}{\theta} - \frac{2n_3 + n_2}{1-\theta} = 0 \Rightarrow \frac{2n_1 + n_2 - \theta(2n_1 + n_2)}{\theta(1-\theta)} = 0$$

$$\Rightarrow 2n_1 + n_2 = \theta(2n) \Rightarrow \theta = \frac{2n_1 + n_2}{2n}$$

T19 (1) 当  $0 < x < \theta$  时,  $F(x) = \int_{-\infty}^x f(x) dx = \int_0^x \frac{3x^2}{\theta^3} dx = \frac{x^3}{\theta^3}$

$$F(x) = \begin{cases} 0, & x \leq 0. \\ \frac{x^3}{\theta^3}, & 0 < x < \theta. \\ 1, & x \geq \theta. \end{cases}$$

$$F_T(t) = P(\max(X_1, X_2, X_3) \leq t) = P(X_1 \leq t, X_2 \leq t, X_3 \leq t),$$

$X_1, X_2, X_3$  独立, 故  $F_T(t) = P(X_1 \leq t)P(X_2 \leq t)P(X_3 \leq t) = [F(t)]^3$

$$= \begin{cases} 0, & t \leq 0 \\ \frac{t^9}{\theta^9}, & 0 < t < \theta \\ 1, & t \geq \theta \end{cases}$$

故  $f_T(t) = \begin{cases} \frac{9t^8}{\theta^9}, & 0 < t < \theta \\ 0, & \text{其他} \end{cases}$

$$(2) E(aT) = aET = a \int_0^\theta t \frac{9t^8}{\theta^9} dt = \frac{9a}{\theta^9} \int_0^\theta t^9 dt = \frac{9a}{\theta^9} \cdot \frac{\theta^{10}}{10} = \frac{9a}{10} \theta = \theta$$

故  $a = \frac{10}{9}$ .