

18. 解: $\because X_i \sim N(0, 0.3^2)$, 且 X_1, X_2, \dots, X_{10} 相互独立

$\therefore Y_i = \frac{X_i}{0.3} \sim N(0, 1)$, 且 Y_1, Y_2, \dots, Y_{10} 相互独立

$$\therefore \sum_{i=1}^{10} Y_i^2 = \frac{1}{0.3^2} \sum_{i=1}^{10} X_i^2 \sim \chi^2(10)$$

$$\therefore P\left\{\sum_{i=1}^{10} X_i^2 > 1.44\right\} = P\left\{\frac{1}{0.3^2} \sum_{i=1}^{10} X_i^2 > 16\right\} = 1 - P\left\{\frac{1}{0.3^2} \sum_{i=1}^{10} X_i^2 \leq 16\right\} = 0.1$$

19. 解: $\because X_i \sim N(\mu, \sigma^2)$, $i=1, 2, \dots, n$, 且各 X_i 相互独立

$$\therefore \frac{X_i - \mu}{\sigma} \sim N(0, 1), i=1, 2, \dots, n$$

$$\therefore \frac{1}{\sigma^2} \sum_{i=1}^n (X_i - \mu)^2 \sim \chi^2(n)$$

$$\text{令 } Y = \sum_{i=1}^n (X_i - \mu)^2 = \sigma^2 \cdot \frac{1}{\sigma^2} \sum_{i=1}^n (X_i - \mu)^2 \stackrel{\sim}{=} \frac{Y}{\sigma^2} \sim \chi^2(n)$$

$$\Rightarrow f_Y(y) = \frac{1}{\sigma^2} f_{\chi^2}\left(\frac{y}{\sigma^2}\right) = \begin{cases} \frac{1}{2^{\frac{n}{2}} \sigma^2 \Gamma(\frac{n}{2})} e^{-\frac{y}{2\sigma^2}} & y > 0 \\ 0 & y \leq 0 \end{cases}$$

30. 解: $X_i \sim N(0, 1)$, $i=1, 2, \dots, 6$ 且相互独立

$$\therefore X_1 + X_2 + X_3 \sim N(0, 3)$$

$$X_4 + X_5 + X_6 \sim N(0, 3)$$

$$\Rightarrow \frac{X_1 + X_2 + X_3}{\sqrt{3}} \sim N(0, 1)$$

$$\frac{X_4 + X_5 + X_6}{\sqrt{3}} \sim N(0, 1)$$

$$\therefore \left(\frac{X_1 + X_2 + X_3}{\sqrt{3}}\right)^2 + \left(\frac{X_4 + X_5 + X_6}{\sqrt{3}}\right)^2 \sim \chi^2(2)$$

$$\Rightarrow C = \frac{1}{3}$$

$$2. \therefore \sum_{i=1}^m x_i \sim N(0, m\sigma^2) \quad \therefore \frac{\sum_{i=1}^m x_i}{\sqrt{m}\sigma} \sim N(0, 1)$$

$$2. \therefore \frac{x_i}{\sigma} \sim N(0, 1) \quad i = m+1, \dots, n$$

$$\frac{x_{m+1}}{\sigma}, \frac{x_{m+2}}{\sigma}, \dots, \frac{x_n}{\sigma} \text{ 相互独立}$$

$$\frac{1}{\sigma} \sum_{i=m+1}^n x_i \sim N(0, n-m)$$

$$\frac{\frac{\sum_{i=1}^m x_i}{\sqrt{m}\sigma}}{\sqrt{\frac{\frac{1}{\sigma^2} \sum_{i=m+1}^n x_i^2}{n-m}}} = \frac{\frac{\sum_{i=1}^m x_i}{\sqrt{m}}}{\sqrt{\sum_{i=m+1}^n x_i^2}} \sim t(n-m)$$

$$\therefore t = \frac{\sum_{i=1}^m x_i}{\sqrt{\sum_{i=m+1}^n x_i^2}}$$

42. 证明

$$t \sim t(n) \Rightarrow t \sim F(1, n)$$

$$\therefore P\{t^2 \leq F_{1-\alpha}(1, n)\} = 1 - \alpha$$

$$\Rightarrow P\{-\sqrt{F_{1-\alpha}(1, n)} \leq t \leq \sqrt{F_{1-\alpha}(1, n)}\} = 1 - \alpha$$

$$\& P\{t \leq -\sqrt{F_{1-\alpha}(1, n)}\} = 1 - P\{t \leq \sqrt{F_{1-\alpha}(1, n)}\}$$

$$\therefore P\{t \leq \sqrt{F_{1-\alpha}(1, n)}\} = 1 - \frac{\alpha}{2}$$

$$\Rightarrow [t_{1-\frac{\alpha}{2}}(n)]^2 = F_{1-\alpha}(1, n)$$

7. 矩估计:

$$f(x, \rho) = \begin{cases} \rho x^{-\rho-1}, & x > 1 \\ 0, & x \leq 1 \end{cases} \Rightarrow \mu = E(X) = \rho \int_1^{+\infty} x \cdot x^{-\rho-1} dx = \frac{\rho}{\rho-1}$$

$$\therefore \hat{\rho} = \frac{\bar{x}}{\bar{x}-1}$$

极大似然估计

$$L(\rho) = \prod_{i=1}^n \rho x_i^{-\rho-1} I_{\{x_i > 1\}}(x_i) \quad (x_1, x_2, \dots, x_n \text{ 均} > 1)$$

$$\Rightarrow \ln L(\rho) = n \ln \rho - (\rho+1) \sum_{i=1}^n \ln x_i$$

$$\frac{d \ln L(\rho)}{d \rho} = \frac{n}{\rho} - \sum_{i=1}^n \ln x_i = 0 \Rightarrow \rho = \frac{n}{\sum_{i=1}^n \ln x_i}$$

$$\textcircled{1} \frac{n}{\sum_{i=1}^n \ln x_i} > 1 \text{ 时 } \hat{\rho} = \frac{n}{\sum_{i=1}^n \ln x_i}$$

$$\textcircled{2} \frac{n}{\sum_{i=1}^n \ln x_i} \leq 1 \text{ 时 由 } \rho > 1 \text{ 及 } \sum_{i=1}^n \ln x_i \geq n \Rightarrow \frac{d \ln L(\rho)}{d \rho} < 0$$

即 $\ln L(\rho)$ 为 ρ ($\rho > 1$) 的单调减函数 $\hat{\rho} = 1$

$$\Rightarrow \hat{\rho} = \max \left\{ 1, \frac{n}{\sum_{i=1}^n \ln x_i} \right\}$$

$$\text{10. (1) } L(\rho) = \rho^n e^{-\rho(\bar{x}-d)} \Rightarrow \ln L(\rho) = n \ln \rho - \rho(\bar{x}-d) \quad (x_1, \dots, x_n > d)$$

$$\frac{d \ln L(\rho)}{d \rho} = \frac{n}{\rho} - (\bar{x}-d) \Rightarrow \hat{\rho} = \frac{1}{\bar{x}-d}$$

$$\text{(2) } L(d) = \prod_{i=1}^n \rho I_{\{x_i > d\}}(x_i) e^{-\rho(x_i-d)} \text{ 为 } d \text{ 的单调减函数 } \Rightarrow \hat{d} = x_{(1)}$$

(3) 固定 ρ , $L(d, \rho)$ 是 d 的单调减函数 $\Rightarrow \hat{d} = x_{(1)}$

$$L(\hat{d}, \rho) \text{ 看成 } \rho \text{ 的函数 } \Rightarrow \hat{\rho} = \frac{1}{\bar{x} - x_{(1)}} \quad \text{故 } \left. \begin{array}{l} \hat{d} = x_{(1)} \\ \hat{\rho} = \frac{1}{\bar{x} - x_{(1)}} \end{array} \right\}$$

13. 解

似然函数

$$L(\mu_1, \mu_2, \sigma^2) = (2\pi)^{-\frac{n_1+n_2}{2}} (\sigma^2)^{-\frac{n_1+n_2}{2}} e^{-\frac{1}{2\sigma^2} [\sum_{i=1}^{n_1} (x_i - \mu_1)^2 + \sum_{i=1}^{n_2} (y_i - \mu_2)^2]}$$

$$\ln L(\mu_1, \mu_2, \sigma^2) = -\frac{n_1+n_2}{2} \ln 2\pi - \frac{n_1+n_2}{2} \ln \sigma^2 - \frac{1}{2\sigma^2} [\sum_{i=1}^{n_1} (x_i - \mu_1)^2 + \sum_{i=1}^{n_2} (y_i - \mu_2)^2]$$

$$\Rightarrow \left\{ \begin{aligned} \frac{\partial \ln L}{\partial \mu_1} &= \frac{n_1}{\sigma^2} (\bar{x} - \mu_1) \\ \frac{\partial \ln L}{\partial \mu_2} &= \frac{n_2}{\sigma^2} (\bar{y} - \mu_2) \end{aligned} \right.$$

$$\frac{\partial \ln L}{\partial \mu_1} = \frac{n_1}{\sigma^2} (\bar{x} - \mu_1)$$

$$\frac{\partial \ln L}{\partial \mu_2} = \frac{n_2}{\sigma^2} (\bar{y} - \mu_2)$$

$$\frac{\partial \ln L}{\partial \sigma^2} = -\frac{n_1+n_2}{2\sigma^2} + \frac{1}{2\sigma^4} [\sum_{i=1}^{n_1} (x_i - \mu_1)^2 + \sum_{i=1}^{n_2} (y_i - \mu_2)^2]$$

$$\text{令 } \frac{\partial \ln L}{\partial \mu_1} = 0, \frac{\partial \ln L}{\partial \mu_2} = 0, \frac{\partial \ln L}{\partial \sigma^2} = 0$$

$$\Rightarrow \hat{\mu}_1 = \bar{x}, \hat{\mu}_2 = \bar{y}, \hat{\sigma}^2 = \frac{1}{n_1+n_2} [\sum_{i=1}^{n_1} (x_i - \bar{x})^2 + \sum_{i=1}^{n_2} (y_i - \bar{y})^2]$$

48. 证明

$$E(T_n) = \frac{2}{n(n+1)} \sum_{i=1}^n i E(X_i) = \mu$$

$$\text{Var}(T_n) = \frac{4}{n^2(n+1)^2} \sum_{i=1}^n i^2 \text{Var}(X_i)$$

$$= \frac{4\sigma^2}{n^2(n+1)^2} \sum_{i=1}^n i^2 = \frac{4\sigma^2}{n^2(n+1)^2} \cdot \frac{n(n+1)(2n+1)}{6} = \frac{2(2n+1)}{3n(n+1)} \sigma^2$$

由 Chebyshev 不等式

$$\text{对 } \forall \epsilon > 0 \text{ 有 } P\{|T_n - \mu| \geq \epsilon\} \leq \frac{\text{Var}(T_n)}{\epsilon^2} = \frac{2(2n+1)}{3n(n+1)} \sigma^2$$

$$\lim_{n \rightarrow \infty} P\{|T_n - \mu| \geq \epsilon\} = 0 \Rightarrow T_n \text{ 是 } \mu \text{ 的相合估计}$$