# 概统作业 (Week 12)

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1

选B

 $(X_i - \mu) \sim N(0,1) \ \Rightarrow \ \sum_{i=1}^n \left(X_i - \mu\right)^2 \sim \chi_n^2$ 

(B)  $(X_n-X_1)\sim N(0,2) \ \Rightarrow \ \frac{\left(X_n-X_1\right)^2}{2}\sim \chi_1^2 \ \Rightarrow \ 2(X_n-X_1)^2$  不服从 $\chi^2$ 分布

(C) 
$$\sum_{i=1}^n \left(X_n - \overline{X}\right)^2 = (n-1)S^2 \sim \chi_{n-1}^2$$

(D) 
$$\sqrt{n}(\overline{X} - \mu) \sim N(0, 1) \implies n(\overline{X} - \mu)^2 \sim \chi_1^2$$

### 2 (P205 T15)

$$\left\{ \begin{array}{ll} E(X_1-2X_2)=0, & Var(X_1-2X_2)=20 \\ E(3X_3-4X_4)=0, & Var(3X_3-4X_4)=100 \end{array} \right. \Rightarrow \left\{ \begin{array}{ll} X_1-2X_2 & \sim N(0,20) \\ 3X_3-4X_4 & \sim N(0,100) \end{array} \right.$$

要使 T 服从  $\chi^2$  分布,有

$$\begin{cases} \sqrt{a}(X_1 - 2X_2) & \sim N(0,1) \\ \sqrt{b}(3X_3 - 4X_4) & \sim N(0,1) \end{cases} \ \Rightarrow \ 20a = 100b = 1 \ \Rightarrow \ a = \frac{1}{20}, \ b = \frac{1}{100}$$

#### 3 (P205 T16)

Proof.

服从自由度为 2 的 t 分布. 不妨设  $X_i \sim N(\mu, \sigma^2)$   $(1 \le i \le 9)$ ,有

$$E(Y_1)=E(Y_2)=\mu, \quad Var(Y_1)=\frac{\sigma^2}{6}, \quad Var(Y_2)=\frac{\sigma^2}{3}$$

故

$$E(Y_1 - Y_2) = 0, \quad Var(Y_1 - Y_2) = \frac{\sigma^2}{2}, \quad Y_1 - Y_2 \sim N(0, \frac{\sigma^2}{2})$$

记 
$$M = \frac{\sqrt{2}}{\sigma}(Y_1 - Y_2)$$
,则 
$$M \sim N(0,1)$$

记  $N=\frac{2S^2}{\sigma^2}$ ,由定理可得

$$N = \frac{2S^2}{\sigma^2} \sim \chi_2^2$$

则

$$Z = rac{\sqrt{2}(Y_1 - Y_2)}{S} = rac{M}{\sqrt{N/2}} \sim t_2.$$

4 (P232 T8)

(1)

$$\begin{split} E(X) &= \int_{-\infty}^{+\infty} x \cdot f(x) dx = \int_{0}^{\theta} \frac{x}{2\theta} dx + \int_{\theta}^{1} \frac{x}{2(1-\theta)} dx = \frac{1}{4} + \frac{\theta}{2} \\ &\overline{X} = \frac{1}{4} + \frac{\hat{\theta}}{2} \ \Rightarrow \ \hat{\theta} = 2\overline{X} - \frac{1}{2}. \end{split}$$

有

(2) 不是。理由如下

$$E(4\overline{X}^2) = 4\left[Var(\overline{X}) + \left(E(\overline{X})\right)^2\right] = 4\left[\frac{Var(X)}{n} + \left(E(X)\right)^2\right]$$

又

$$E(X) = \frac{1}{4} + \frac{\theta}{2}, \quad E(X^2) = \int_{-\infty}^{+\infty} x^2 f(x) dx = \frac{2\theta^2 + \theta + 1}{6}$$

故

$$Var(X) = E(X^2) - E(X)^2 = \frac{2\theta^2 + \theta + 1}{6} - \left(\frac{1}{4} + \frac{\theta}{2}\right)^2 = \frac{\theta^2}{12} - \frac{\theta}{12} + \frac{5}{48}.$$

原式即为

$$E(4\overline{X}^2) = 4\left[\left(\frac{\theta^2}{12n} - \frac{\theta}{12n} + \frac{5}{48n}\right) + \left(\frac{\theta^2}{4} + \frac{\theta}{4} + \frac{1}{16}\right)\right] = \frac{3n+1}{3n}\theta^2 + \frac{3n-1}{3n}\theta + \frac{3n+5}{12} \neq \theta^2$$

即证  $4\overline{X}^2$  不是  $\theta^2$  的无偏估计量.

## 5 (P234 T27)

有最大似然函数

$$\begin{split} L(X_1,X_2,\dots,X_n;\lambda) &= \prod_{i=1}^n \frac{\lambda^{x_i}}{X_i!} e^{-\lambda} = e^{-n\lambda} \prod_{i=1}^n \frac{\lambda^{X_i}}{X_i!}.\\ \ln L &= -n\lambda + \sum_{i=1}^n \left(X_i \ln \lambda - \ln X!\right) \end{split}$$

令

$$\frac{\partial \ln L}{\partial \lambda} = \frac{1}{\lambda} \sum_{i=1}^n X_i - n = 0.$$

得

$$\hat{\lambda} = \frac{1}{n} \sum_{i=1}^{n} X_i = \overline{X}$$

检验二阶导

$$\frac{\partial^2 \ln L}{\partial \lambda^2} = -\frac{1}{\lambda^2} \sum_{i=1}^n X_i$$

对于所有  $\lambda$  与  $\overline{X}$  非零时为负,故所求驻点为 L 极大值点, P(X=0) 的最大似然估计为

$$P(X=0) = e^{-\widehat{\lambda}} = e^{-\overline{X}}$$

#### 6 (P235 T29)

(1) 即  $X \sim U(\theta, 0), \theta \in \Theta$ , 有矩估计

$$E(X) = \frac{0+\theta}{2} = \overline{X} \ \Rightarrow \ \hat{\theta} = 2\overline{X}$$

有最大似然函数

$$L(X_1,X_2,\dots,X_n;\theta) = \prod_{i=1}^n P(X=X_i) = \begin{cases} \frac{1}{(-\theta)^n}, & X_1,X_2,\dots,X_n \in [\theta,0] \\ 0. & \sharp \text{ id} \end{cases}$$

显然当  $(-\theta)^n$  最小,即  $\theta$  最大时,L 最大,又

$$\theta \leq \min\{X_1, X_2, \dots, X_n\}, \ \theta \in (-\infty, 0)$$

有最大似然估计

$$\hat{\theta} = \min\{X_1, X_2, \dots, X_n\}.$$

(2) 即  $X \sim U(\theta, 2\theta), \theta \in \Theta$ , 有矩估计

$$E(X) = \frac{\theta + 2\theta}{2} = \overline{X} \ \Rightarrow \ \hat{\theta} = \frac{2}{3}\overline{X}$$

有最大似然函数

$$L(X_1,X_2,\dots,X_n;\theta)=\prod_{i=1}^n P(X=X_i)=\begin{cases} \frac{1}{\theta^n}, & X_1,X_2,\dots,X_n\in[\theta,2\theta]\\ 0. & \not\exists \text{ th} \end{cases}$$

显然当  $\theta^n$  最小时, L 最大, 又

$$\theta \le \min\{X_1, X_2, \dots, X_n\}, \ \theta \ge \max\{X_1, X_2, \dots, X_n\}, \ \theta \in (0, +\infty)$$

有最大似然估计

$$\hat{\theta} = \frac{1}{2} \max\{X_1, X_2, \dots, X_n\}.$$