

## 习题 6.2

1. (1)  $\hat{\theta} = 2\bar{X} - \frac{1}{2}$ ; (2) 因为  $E(4\bar{X}^2) > \theta^2$ , 所以  $4\bar{X}^2$  不是  $\theta^2$  的无偏估计.

2. 解:  $X \sim E(\theta)$ ,  $E(\bar{X}) = E(X) = \theta$ , 所以  $\bar{X}$  是  $\theta$  的无偏估计.

令  $T = \min\{X_1, X_2, \dots, X_n\}$ ,

$T$  的分布函数  $F_T(t) = P(T \leq t)$

$$\begin{aligned} &= 1 - P(T > t) \\ &= 1 - P(X_1 > t, X_2 > t, \dots, X_n > t) \\ &= 1 - [P(X > t)]^n \\ &= 1 - [1 - P(X \leq t)]^n \\ &= 1 - [1 - F_X(t)]^n \\ &= 1 - e^{-\frac{nt}{\theta}}, \quad t > 0, \end{aligned}$$

$T$  的概率密度  $f_T(t) = (F_T(t))' = \frac{n}{\theta} e^{-\frac{nt}{\theta}}, t > 0$ ,

即  $T \sim E\left(\frac{\theta}{n}\right)$ , 故  $E(T) = \frac{\theta}{n}$ , 从而  $E(nT) = \theta$ ,

所以  $nT$  也是  $\theta$  的无偏估计.

3.  $\mu_2$  最有效.

4.  $\frac{4}{3} \max\{X_1, X_2, X_3\}$  比  $4 \min\{X_1, X_2, X_3\}$  更有效.

解: 令  $M = \max\{X_1, X_2, X_3\}$ ,  $N = \min\{X_1, X_2, X_3\}$ , 先求  $M, N$  的概率密度.

$F_M(x) = P\{M \leq x\} = P\{X_1 \leq x, X_2 \leq x, X_3 \leq x\}$

$$= [P\{X \leq x\}]^3 = \begin{cases} 1, & x \geq \theta, \\ \frac{x^3}{\theta^3}, & 0 < x < \theta, \\ 0, & x \leq 0. \end{cases}$$

$$\therefore f_M(x) = \begin{cases} \frac{3x^2}{\theta^3}, & 0 < x < \theta, \\ 0, & \text{其他.} \end{cases}$$

$F_N(x) = P\{N \leq x\} = 1 - P\{N > x\} = 1 - P\{X_1 \geq x, X_2 \geq x, X_3 \geq x\}$

$$= 1 - [1 - P\{X < x\}]^3 = \begin{cases} 1, & x \geq \theta, \\ 1 - \left(1 - \frac{x}{\theta}\right)^3, & 0 < x < \theta, \\ 0, & x \leq 0. \end{cases}$$

$$\therefore f_N(x) = \begin{cases} \frac{3}{\theta} \left(1 - \frac{x}{\theta}\right)^2, & 0 < x < \theta, \\ 0, & \text{其他.} \end{cases}$$

$$E\left(\frac{4}{3}M\right) = \frac{4}{3} \int_0^\theta \frac{3x^3}{\theta^3} dx = \int_0^\theta \frac{4x^3}{\theta^3} dx = \theta,$$

$$E(4N) = 4 \int_0^\theta \left[ \frac{3}{\theta} \left(1 - \frac{x}{\theta}\right)^2 \right] x dx = \frac{12}{\theta} \int_0^\theta \left(x + \frac{x^3}{\theta^2} - \frac{2x^2}{\theta}\right) dx = \theta,$$

$\therefore \frac{4}{3} \max\{X_1, X_2, X_3\}, 4 \min\{X_1, X_2, X_3\}$  都是  $\theta$  的无偏估计.

$$E(M) = \frac{3}{4}\theta, E(M^2) = \int_0^\theta \frac{3x^4}{\theta^3} dx = \frac{3\theta^2}{5},$$

$$E(N) = \frac{\theta}{4}, E(N^2) = \int_0^\theta \left[ \frac{3}{\theta} \left(1 - \frac{x}{\theta}\right)^2 \right] x^2 dx = \frac{3}{\theta} \int_0^\theta \left(x^2 + \frac{x^4}{\theta^2} - \frac{2x^3}{\theta}\right) dx = \frac{\theta^2}{10},$$

$$\therefore D(M) = E(M^2) - E^2(M) = \frac{3}{80}\theta^2, D(N) = E(N^2) - E^2(N) = \frac{3}{80}\theta^2,$$

$$\therefore D\left(\frac{4}{3}M\right) = \frac{1}{15}\theta^2 < D(N) = \frac{3}{5}\theta^2,$$

$\therefore \frac{4}{3} \max\{X_1, X_2, X_3\}$  比  $4 \min\{X_1, X_2, X_3\}$  更有效.

5. 解:  $X \sim U(0, \theta), D(X) = \sigma^2 = \frac{\theta^2}{12},$

$$\therefore E\left[\frac{12}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2\right] = 12E(S^2) = 12D(X) = 12 \cdot \frac{\theta^2}{12} = \theta^2,$$

$\therefore \frac{12}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$  是  $\theta^2$  的无偏估计.

$$\therefore \frac{(n-1)S^2}{\sigma^2} \sim \chi^2(n-1), D\left[\frac{(n-1)S^2}{\sigma^2}\right] = 2(n-1),$$

$$\therefore D(S^2) = \frac{2\sigma^4}{n-1} = \frac{\theta^4}{72(n-1)}.$$

由切比雪夫不等式,  $P\{|12S^2 - \theta^2| < \varepsilon\} \geq 1 - \frac{D(12S^2)}{\varepsilon^2} = 1 - \frac{144D(S^2)}{\varepsilon^2} = 1 - \frac{2\theta^4}{(n-1)\varepsilon^2}$

$$\lim_{n \rightarrow \infty} P\{|12S^2 - \theta^2| < \varepsilon\} \geq \lim_{n \rightarrow \infty} \left[1 - \frac{2\theta^4}{(n-1)\varepsilon^2}\right] = 1,$$

$\therefore \frac{12}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$  是  $\theta^2$  的相合估计.