

1. 设  $Y_i = (X_i - \mu)^2$ , 则  $Y_i$  iid

$$E Y_i = E (X_i - \mu)^2 = \sigma^2$$

由 Khintchine:

$$\frac{1}{n} \sum Y_i \xrightarrow{P} \sigma^2$$

$$\text{即 } \frac{1}{n} \sum_{k=1}^n (X_k - \mu)^2 \xrightarrow{P} \sigma^2$$

$$2. D\left(\frac{1}{n} \sum_{k=1}^n X_k\right) = \frac{1}{n^2} D\left(\sum_{k=1}^n X_k\right)$$

$$= \frac{1}{n^2} \left[ \sum_{k=1}^n D X_k + \sum_{i \neq j} \text{Cov}(X_i, X_j) \right]$$

$$\text{由 Cauchy: } \text{Cov}(X_i, X_j) \leq \sqrt{D X_i D X_j} \quad \therefore \text{Cov}(X_i, X_j) \leq C$$

$$D\left(\frac{1}{n} \sum_{k=1}^n X_k\right) = \frac{1}{n^2} \left[ \sum_{k=1}^n D X_k + \sum_{k=1}^n \sum_{i \neq j} \text{Cov}(X_i, X_j) + \sum_{i=1}^n \sum_{j \neq i} \text{Cov}(X_i, X_j) \right]$$

$$\leq \frac{1}{n^2} [nC + 2N_n \cdot C + n^2 \cdot \varepsilon_0 \cdot C]$$

$$P\left(\left|\frac{1}{n} \sum_{k=1}^n X_k - \frac{1}{n} \sum_{k=1}^n E X_k\right| \geq \varepsilon\right) \leq \frac{D \frac{1}{n} \sum_{k=1}^n X_k}{\varepsilon^2}$$

$$\leq \frac{N(1+2n)C}{n^2 \varepsilon^2} + \frac{\varepsilon_0 C}{\varepsilon^2}$$

$$\text{取 } \varepsilon_0 = \frac{1}{n}$$

$$\text{故原式} \leq \frac{N(1+2n)C}{n^2 \varepsilon^2} + \frac{C}{n^2 \varepsilon^2}$$

$\therefore$  在  $n \rightarrow +\infty$  时

原式  $\rightarrow 0$

3. 设  $Y_k = \cos kx$

$$E Y_k = \int_{-\pi}^{\pi} \cos kx \cdot \frac{1}{2\pi} dx = 0$$

$i \neq j$ :

$$\text{Cov}(Y_i, Y_j) = E Y_i Y_j - E Y_i E Y_j$$

$$= E Y_i Y_j = \int_{-\pi}^{\pi} \cos i x \cos j x dx = 0$$

$$D Y_k = E Y_k^2 = \int_{-\pi}^{\pi} (\cos kx)^2 dx = \int_{-\pi}^{\pi} \frac{1 + \cos 2kx}{2} dx = \frac{1}{2}$$

$\therefore \{Y_k\}$  两两不相关且方差一致有界

由 Chebyshev 大数定律:

$$\frac{S_n}{n} - E\left(\frac{S_n}{n}\right) = \frac{S_n}{n} \Rightarrow 0$$

$$4. \ln Y_n = \frac{\sum_{k=1}^n \ln X_k}{n}$$

$$\text{令 } Z_n = \ln X_n$$

$$P(Z_n \leq z) = P(\ln X_n \leq z)$$

$$= P(X_n \leq e^z) = \int_0^{e^z} 1 dx = e^z \quad (z \leq 0)$$

$$f_{Z_n}(z) = (e^z)' = e^z \quad (z \leq 0)$$

$$E Z_n = \int_{-\infty}^0 z e^z dz = -1$$

$\{Z_n\}$  独立同分布, 由 Khintchine 大数定律:

$$\frac{\sum_{k=1}^n Z_k}{n} \xrightarrow{P} -1$$

故  $\ln Y_n \xrightarrow{P} -1$  而  $e^x$  为连续函数

$$\text{故 } e^{\ln Y_n} \xrightarrow{P} e^{-1}$$

$$\text{即 } Y_n \xrightarrow{P} \frac{1}{e} \quad \therefore c = \frac{1}{e}$$

$$5. \text{ 设 } A_i = \begin{cases} 1 & \text{第 } i \text{ 次试验正面朝上} \\ 0 & \text{第 } i \text{ 次试验背面朝上} \end{cases}$$

$$E A_i = 0.5$$

$$D A_i = 0.25$$

则  $X = \frac{\sum_{i=1}^n A_i}{n}$  为正面朝上的频率

$$E X = \frac{\sum_{i=1}^n E A_i}{n} = 0.5$$

$$D X = \frac{1}{4n}$$

$$P(0.4 < X < 0.6) = P(|X - EX| < 0.1)$$

由切比雪夫不等式:

$$P(|X - EX| \geq 0.1) \leq \frac{D X}{0.1^2} = \frac{\frac{1}{4n}}{0.1^2} = \frac{25}{n}$$

$$P(|X - EX| < 0.1) = 1 - P(|X - EX| \geq 0.1)$$

$$\geq 1 - \frac{25}{n} > 0.9$$

故  $n \geq 250$

用正态近似估计:

$$P(0.4 < \frac{\sum A_i}{n} < 0.6) = P(\frac{0.4-0.5}{\frac{0.5}{\sqrt{n}}} < \frac{\frac{\sum A_i}{n} - 0.5}{\frac{0.5}{\sqrt{n}}} < \frac{0.6-0.5}{\frac{0.5}{\sqrt{n}}})$$

$$= P(-\frac{\sqrt{n}}{5} < Z < \frac{\sqrt{n}}{5}) > 0.9$$

查表得:  $P(Z \leq 1.65) = 0.9509$

$$\frac{\sqrt{n}}{5} \geq 1.65$$

$$n \geq 68$$

6. (a) 设  $A_i = \begin{cases} 1 & \text{右移} \\ -1 & \text{左移} \end{cases}$   $A_i$  独立同分布  $\begin{cases} DA_i = 1 \\ EA_i = 0 \end{cases}$

$$X_n = \sum_{i=1}^n A_i$$

$$DX_n = D(\sum_{i=1}^n A_i) = \sum_{i=1}^n DA_i = n$$

(b)

$$S_n^* = \frac{\sum_{i=1}^n A_i - n \cdot EA_i}{\sqrt{n \cdot DA_i}} = \frac{\sum_{i=1}^n A_i}{\sqrt{n}} = \frac{X_n}{\sqrt{n}}$$

由 Lindeberg-Levy 中心极限定理

$$S_n^* \xrightarrow{D} N(0, 1)$$

$$\text{即 } \lim_{n \rightarrow \infty} F_n(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{t^2}{2}} dt$$

$$\text{故 } P(\frac{X_n}{\sqrt{n}} \leq x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{t^2}{2}} dt$$

7.  $X_i \stackrel{iid}{\sim} U(-1, 1)$

$$\therefore EX_i = 0 \quad DX_i = \frac{1}{3}$$

$$\text{则 } E(\bar{x}) = E(\frac{\sum X_i}{n}) = \frac{1}{n} \sum EX_i = 0$$

$$D(\bar{x}) = D(\frac{\sum X_i}{n}) = \frac{1}{n^2} \sum DX_i = \frac{1}{3n}$$

8.  $Cov(X_i - \bar{x}, X_j - \bar{x}) = Cov(X_i, X_j) - Cov(X_i, \bar{x}) - Cov(\bar{x}, X_j) + Cov(\bar{x}, \bar{x})$

$$= 0 - \frac{1}{n} DX_i - \frac{1}{n} DX_j + \frac{1}{n^2} \sum DX_i$$

设  $\forall i, DX_i = \sigma^2$  故  $Cov(X_i - \bar{x}, X_j - \bar{x}) = -\frac{\sigma^2}{n}$

$$D(X_i - \bar{x}) = D(X_j - \bar{x}) = D(X_i - \bar{x}) = D(\frac{(1-i)X_1 - X_2 - \dots - X_n}{n}) = \frac{n-1}{n} \sigma^2$$

$$\text{Cov}(X_i - \bar{X}, X_j - \bar{X}) = \frac{\text{Cov}(X_i - \bar{X}, X_j - \bar{X})}{\sqrt{D_{X_i - \bar{X}} D_{X_j - \bar{X}}}} = -(n-1)^{-1}$$

$$9. \quad X_i \sim N(8, 4) \quad Y_i = \frac{X_i - 8}{2} \sim N(0, 1)$$

$$(1) \quad P(X_{(n)} > 10) = 1 - P(X_{(n)} \leq 10) = 1 - [P(X_1 \leq 10)]^n = 0.9370$$

$$(2) \quad P(X_{(1)} > 5) = [P(X_i > 5)]^n = 0.3308$$

$$10. \quad \text{令 } Z_i = X(i)$$

$$\therefore \begin{cases} Z_1 = y_1 y_2 \cdots y_n \\ Z_2 = y_2 y_3 \cdots y_n \\ \vdots \\ Z_n = y_n \end{cases}$$

$$\therefore \text{Jacob 行列式为 } |J| = y_2 y_3^2 y_4^3 \cdots y_n^{n-1}$$

$$\begin{aligned} f_{Y_1, Y_2, \dots, Y_n}(y_1, \dots, y_n) &= \int_{Z_1, Z_2, \dots, Z_n} (Z_1(y_1, \dots, y_n), \dots, Z_n(y_1, \dots, y_n)) |J| \\ &= n! \prod_{i=1}^n \frac{1}{\theta^i} \cdot |J| = \frac{n!}{\theta^n} y_2 y_3^2 \cdots y_n^{n-1} \end{aligned}$$

可分离变量  $\therefore y_1, \dots, y_n$  相互独立