

# 第十四周第一次作业:

一. P184-185 3.  $\{z_n\}$  方差有界,  $|i-j| \rightarrow \infty$  时  $z_i$  与  $z_j$  相关系数  $r_{ij} \rightarrow 0$

Prove:  $\{z_n\}$  收敛于大数定律

Proof. 只需证明  $\frac{1}{n} \sum_{i=1}^n z_i \rightarrow E[z_i]$  a.s.

定义样本平均  $\bar{z}_n = \frac{1}{n} \sum_{i=1}^n z_i$

$$\begin{aligned} \text{方差 } D(\bar{z}_n) &= D\left(\frac{1}{n} \sum_{i=1}^n z_i\right) = \frac{1}{n^2} D\left(\sum_{i=1}^n z_i\right) \\ &= \frac{1}{n^2} \left[ \sum_{i=1}^n D(z_i) + 2 \sum_{1 \leq i < j \leq n} \text{cov}(z_i, z_j) \right] \end{aligned}$$

因  $z_i \leq c$ , 故

$$\sum_{i=1}^n D(z_i) \leq nc$$

对协方差项: 记  $\text{cov}(z_i, z_j) = \rho_{ij} \sqrt{D(z_i)D(z_j)}$

$$\Rightarrow |\text{cov}(z_i, z_j)| \leq c$$

$$\text{当 } |i-j| \leq N \text{ 时: } \sum_{\substack{1 \leq i < j \leq n \\ |i-j| \leq N}} \text{cov}(z_i, z_j) \leq nNC$$

$$\text{当 } |i-j| > N \text{ 时: } \sum_{\substack{1 \leq i < j \leq n \\ |i-j| > N}} |\text{cov}(z_i, z_j)| \leq \varepsilon \cdot \sum_{\substack{1 \leq i < j \leq n \\ |i-j| > N}} c = \varepsilon n^2 \frac{c}{2} \left(1 - \frac{N}{n}\right)$$

$$\text{综上有: } D\left(\sum_{i=1}^n z_i\right) \leq nc + 2nNC + \varepsilon c \frac{n^2}{2} \left(1 - \frac{N}{n}\right)$$

$$D(\bar{z}) = \frac{1}{n^2} D\left(\sum_{i=1}^n z_i\right) \leq \frac{c}{n} + \frac{2NC}{n} + \varepsilon c \left(\frac{1}{2} - \frac{N}{2n}\right)$$

with  $n \rightarrow +\infty$  上式右侧为 0

最后由 Chebechev 不等式:  $\forall \delta > 0$

$$P(|\bar{z}_n - \bar{z}| > \frac{1}{n}) \leq \frac{D(\bar{z}_n)}{\frac{1}{n^2}} \rightarrow 0 \quad \frac{1}{n} \rightarrow 0$$

故成立大数定律

4.  $\{z_n\}$  iid r.v.s.  $\forall n, D(z_n) < +\infty$ ,  $|k-l| \geq 2$  时  $z_k$  与  $z_l$  独立

证: 对  $\{z_n\}$  大数定律成立

Proof: 首先构造新的随机变量:  $\{z_{3k+1}\}, \{z_{3k+2}\}, \{z_{3k}\}$

它们每组中均满足独立性, 每个方差为  $\sigma^2$

只需证  $\sum_{n=1}^{\infty} \frac{D(z_n)}{n^2}$  收敛:

因所有  $z_n$  方差相同,  $D(z_n) = \sigma^2$ , 于是:

$$\sum_{n=1}^{\infty} \frac{\sigma^2}{n^2} = \sigma^2 \sum_{n=1}^{\infty} \frac{1}{n^2} = \frac{\pi^2}{6} \sigma^2$$

$$\text{所以 } \sigma^2 \sum_{n=1}^{\infty} \frac{1}{n^2} = \sigma^2 \frac{\pi^2}{6} < +\infty$$

因此  $\sum_{n=1}^{\infty} \frac{D(z_n)}{n^2}$  收敛

5.  $\{z_k\}$  iid  $\sim U(0,1)$ , 求证:  $(\prod_{k=1}^n z_k)^{\frac{1}{n}} \xrightarrow{P} c$ , 并求  $c$

Proof: 令  $Y_n = \log(\prod_{k=1}^n z_k)^{\frac{1}{n}} = \frac{1}{n} \sum_{k=1}^n \log z_k$

因为  $z_k \sim U(0,1)$ ,  $\log z_k$  iid

$$E(\log z_k) = \int_0^1 \log x dx = -1$$

$$E[(\log z_k)^2] = \int_0^1 (\log x)^2 dx = 2$$

$$\Rightarrow D(\log z_k) = 2 - 1 = 1$$

$$\Rightarrow (\prod_{k=1}^n z_k)^{\frac{1}{n}} = e^{Y_n} \xrightarrow{P} \frac{1}{e}, \quad c = \frac{1}{e}$$

8. 举例说明 Borel-Cantelli 引理中, 命题.

$$(1) \text{ 若 } \sum_{n=1}^{\infty} P(A_n) < +\infty, \text{ 则 } P(\overline{\lim}_{n \rightarrow \infty} A_n) = 0$$

的逆不成立

例子: 概率空间  $(\Omega, \mathcal{F}, P)$ , 其中  $\Omega = [0, 1]$ ,  $\mathcal{F}$  为 Borel 代数,

$P$  是区间上的 Lebesgue 测度

$$\text{令 } A_n = [\frac{1}{2^n}, 1]$$

$$\text{则 } P(\overline{\lim}_{n \rightarrow \infty} A_n) = 0, \text{ 但 } \sum_{n=1}^{\infty} P(A_n) = +\infty$$

9.  $\{z_n\}$  iid, 方差有限  $D(z_n) = \sigma^2$ ,  $\bar{z} = E\{z_k, k \geq 1\}$ ,  $S_n^2 = D(z_1, z_2, \dots, z_n)$

试证:  $\{S_n^2\}$  几乎必然收敛到  $\sigma^2$

$$\text{Proof: } \bar{z}_n = \frac{1}{n} \sum_{i=1}^n z_i, S_n^2 = \frac{1}{n-1} \sum_{i=1}^n (z_i - \bar{z}_n)^2$$

先计算  $E(z_i) = \mu$ .  $z_i$  iid, 样本方差:

$$S_n^2 = \frac{1}{n-1} \sum_{i=1}^n (z_i - \frac{1}{n} \sum_{j=1}^n z_j)^2$$

$$\text{注意到 } 2 \bar{z}_n \sum_{i=1}^n z_i = 2 \bar{z}_n \cdot n \bar{z}_n = 2n \bar{z}_n^2$$

$$\Rightarrow S_n^2 = \frac{1}{n-1} \left( \sum_{i=1}^n z_i^2 - n \bar{z}_n^2 \right)$$

$$\text{由大数定律: } \bar{z}_n \xrightarrow{a.s.} \mu$$

$$\text{并且 } \bar{z}_n^2 \xrightarrow{a.s.} \mu^2$$

再由大数定律  $z_i^2$ :

$$\frac{1}{n} \sum_{i=1}^n z_i^2 \sim n(\sigma^2 + \mu^2)$$

$$\Rightarrow S_n^2 \approx \frac{1}{n-1} [n(\sigma^2 + u^2) - n\bar{z}_n^2]$$

由于  $\bar{z}_n \xrightarrow{a.s.} u^2$  故:

$$n(\sigma^2 + u^2) - n\bar{z}_n^2 \xrightarrow{a.s.} n\sigma^2$$

10 设  $\{z_n\}$  为独立随机变量序列, 满足条件  $E(z_n) \rightarrow 0$  且  $\sum_{n=1}^{\infty} \frac{E(z_n^2)}{n^2} < +\infty$

$$\text{证明: } \frac{1}{n} \sum_{k=1}^n z_k \xrightarrow{a.s.} 0$$

Proof: 令  $\eta_n = z_n - E(z_n)$ , 则  $\eta_n$  都独立同分布,  $E(\eta_n) = 0$

$$E(\eta_n^2) = E[(z_n - E(z_n))^2] = E(z_n^2) - E^2(z_n)$$

因  $E(z_n) \rightarrow 0$ , 对足够多的  $n$ ,  $E(z_n)$  较小

$$\text{即 } E(\eta_n^2) \approx E(z_n^2)$$

$$\text{由此可得: } \sum_{n=1}^{\infty} \frac{E(\eta_n^2)}{n^2} \leq \sum_{n=1}^{\infty} \frac{E(z_n^2)}{n^2} < +\infty$$

由 Kolmogorov 强大数定律推广版本:

$$\frac{1}{n} \sum_{k=1}^n z_k = \frac{1}{n} \sum_{k=1}^n [\eta_k + E(z_k)] = \frac{1}{n} \sum_{k=1}^n \eta_k + \frac{1}{n} \sum_{k=1}^n E(z_k)$$

设  $\lim_{n \rightarrow \infty} E(z_k) = 0$ , 则有:

$$\left| \frac{1}{n} \sum_{k=1}^n E(z_k) \right| \leq \frac{1}{n} \sum_{k=1}^n |E(z_k)| \rightarrow 0$$

$$\Rightarrow \frac{1}{n} \sum_{k=1}^n z_k = \frac{1}{n} \sum_{k=1}^n \eta_k + \frac{1}{n} \sum_{k=1}^n E(z_k) \rightarrow 0 \quad a.s.$$

二. P111 2. Theorem 5.1.2:  $X_j$ 's are uncorrelated and their second moments

have a second bound, then (1) is true in  $L^2$  and hence also pr

$$(1) \frac{S_n - E(S_n)}{n} \rightarrow 0 \quad a.e.$$

Under the same hypothesis we have:

$$\frac{S_n}{n^d} \rightarrow 0 \text{ a.e. for any } d > \frac{3}{4} \quad ?$$

Prig 3. For any sequence  $\{x_n\}$ :

$$\frac{S_n}{n} \rightarrow 0 \text{ in pr.} \Rightarrow \frac{x_n}{n} \rightarrow 0 \text{ in pr.}$$

More generally, this is true if  $n$  is replaced by  $b_n$ ,

$$\text{where } \frac{b_{n+1}}{b_n} \rightarrow 1$$

Proof: Assume that  $\frac{S_n}{n} \rightarrow 0$  in probability, which means.

$$\forall \varepsilon > 0, \lim_{n \rightarrow \infty} P\left(\left|\frac{S_n}{n}\right| \geq \varepsilon\right) = 0$$

We need to show that:

$$\forall \varepsilon > 0, \lim_{n \rightarrow \infty} P\left(\left|\frac{x_n}{n}\right| \geq \varepsilon\right) = 0$$

Note that:

$$S_n = \sum_{i=1}^n x_i$$

$$\text{So we have } \frac{S_n}{n} = \frac{1}{n} \sum_{i=1}^n x_i$$

For any  $\varepsilon > 0$ .

$$P\left(\left|\frac{x_n}{n}\right| \geq \varepsilon\right) \leq P\left(\left|\frac{S_n}{n}\right| \geq \varepsilon\right) + P\left(\left|\frac{S_n - x_n}{n}\right| \geq \varepsilon\right)$$

As  $n \rightarrow \infty$ , since  $\frac{S_n}{n} \rightarrow 0$  in p. the first term on the right-hand side goes to 0.

$$\lim_{n \rightarrow \infty} P(|\frac{S_n}{n}| \geq \varepsilon) = 0$$

For the second term, observe that:

$$\frac{S_n - X_n}{n} = \frac{1}{n} \sum_{i=1}^{n-1} X_i$$

which is essentially the average of  $X_1, X_2, \dots, X_{n-1}$

which also should go to 0 in probability due to the

same reasoning applied to  $S_n$

Thus the second term goes to 0:

$$\lim_{n \rightarrow \infty} P(|\frac{S_n}{n}| \geq \varepsilon) = 0$$

Combining these results we get:

$$\lim_{n \rightarrow \infty} P(|\frac{S_n}{n}| \geq \varepsilon) = 0$$

This shows that  $\frac{X_n}{n} \rightarrow 0$  in probability

Generalization with  $b_n$ :

For a sequence  $\{b_n\}$  such that  $\frac{b_{n+1}}{b_n} \rightarrow 1$

Use the same arguments as before, we can show that

$\frac{X_n}{b_n} \rightarrow 0$  in probability