第五章、极限定理 §5.1 Bernoulli 试验场合的极限定理

• 假设 $X = X_1, X_2, \cdots$ 独立同分布(i.i.d.),

$$P(X = 1) = p, \quad P(X = 0) = q = 1 - p.$$

 $S_n = X_1 + \dots + X_n.$

• $P(S_n = k) = C_n^k p^k q^{n-k}, k = 0, 1, \dots, n.$

$$ES_n = np$$
, $var(S_n) = npq$.

$$E\xi_n = \mathbf{p}, \quad \operatorname{var}(\xi_n) = \frac{n \cdot \operatorname{var}(X)}{n} = \frac{1}{n} \operatorname{var}(X) = \frac{1}{n} pq.$$

• (1713年) Bernoulli 大数定律(5.1.14):

$$P\left(\left|\frac{S_n}{n} - p\right| \geqslant \varepsilon\right) \leqslant \frac{1}{\varepsilon^2} \operatorname{var}\left(\frac{S_n}{n}\right) = \frac{1}{\varepsilon^2} \cdot \frac{\operatorname{var}(X)}{n} \to 0.$$

如果∀ε > 0 都有

$$\lim_{n\to\infty} P(|\xi_n - \xi| \geqslant \varepsilon) = 0.$$

那么, 称 ξ_n 依概率收敛到 ξ , 记为 $\xi_n \stackrel{P}{\to} \xi$. (定义5.2.3)

•
$$S_n^* = \frac{S_n - np}{\sqrt{npq}}$$
: $x = \varepsilon/\sqrt{pq}$,

$$P\left(\left|\frac{S_n - np}{\sqrt{n}}\right| \geqslant \varepsilon\right) = P\left(\left|S_n^*\right| \geqslant x\right) \leqslant \frac{1}{x^2}, \quad \forall x > 0.$$

• (1733, 1778) De Moivre-Laplace 中心极限定理(定理5.1.1):

$$P(a < S_n^* \leqslant b) \to \int_a^b \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx, \quad \forall a < b.$$

•
$$x_k = \frac{k - np}{\sqrt{npq}}$$
, $\mathbb{K} = \left(np + a\sqrt{npq}, np + b\sqrt{npq}\right]$, $\Delta x = \frac{1}{\sqrt{npq}}$.

$$\frac{P(S_n = k)}{p_Z(x_k)\Delta x} = \frac{P(S_n = k)}{\frac{1}{\sqrt{2\pi}}e^{-\frac{1}{2}x_k^2}\frac{1}{\sqrt{npq}}} \stackrel{\mathbb{K}}{\Rightarrow} 1.$$

• 进一步,

$$\lim_{n} \frac{\sum_{k \in \mathbb{K}} P(S_n = k)}{\int_a^b \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx} = \lim_{n} \frac{\sum_{k \in \mathbb{K}} P(S_n = k)}{\sum_{k \in \mathbb{K}} p_Z(x_k) \Delta x} = 1.$$

• a 可为 $-\infty$; b 可为 ∞ .

$$P(S_n^* \le x) \to \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz = \Phi(x), \quad \forall x \in \mathbb{R}.$$

• 假设 $\xi \sim N(0,1)$. 如果 $\forall x$ 都有

$$\lim_{n\to\infty} P(\xi_n \leqslant x) = P(\xi \leqslant x).$$

那么, 称 ξ_n 依分布收敛于 ξ , 记为 $\xi_n \stackrel{d}{\to} \xi$.

§5.2 之依概率收敛, §5.1 & §5.3 之弱大数定律

• 如果 $\forall \varepsilon > 0$ 都有

$$\lim_{n\to\infty} P(|\xi_n - \xi| \ge \varepsilon) = 0.$$

那么, 称 ξ_n 依概率收敛到 ξ , 记为 $\xi_n \stackrel{P}{\to} \xi$. (定义5.2.3)

如果

$$\lim_{n\to\infty} E|\xi_n - \xi|^r = 0.$$

那么, 称 ξ_n r 阶(平均)收敛到 ξ , 记为 $\xi_n \stackrel{r}{\to} \xi$. (定义5.2.4)

•
$$\xi_n \xrightarrow{r} \xi \Rightarrow \xi_n \xrightarrow{P} \xi$$
. (定理5.2.8)

$$P(|\xi_n - \xi| \ge \varepsilon) \le \frac{1}{\varepsilon^r} E|\xi_n - \xi|^r.$$

• "
$$\xi_n \xrightarrow{P} \xi \Rightarrow \xi_n \xrightarrow{\tau} \xi$$
" 不成立. (例5.2.4)
$$\Omega = (0,1], \ \Pi \cap \mathbb{R} \mathbb{Z}.$$

$$\xi_n = n^{\frac{1}{r}} \times 1_{(0,\frac{1}{n}]},$$

 $\xi \equiv 0.$

•
$$\xi_n \stackrel{P}{\to} \xi \& E|\xi_n|^r \to E|\xi|^r \Rightarrow \xi_n \stackrel{r}{\to} \xi.$$

(证明不要求)



弱大数定律(Weak Law of Large Numbers, WLLN)形式:
 若X₁, X₂, · · · 满足 ***, 则

$$\frac{S_n}{b_n} - a_n \stackrel{P}{\to} 0.$$

• WLLN 的证明方法:

$$P\left(\left|\frac{S_n}{b_n} - a_n\right| \ge \varepsilon\right) \le \frac{1}{\varepsilon^r} E\left|\frac{S_n}{b_n} - a_n\right|^r, \quad r > 0.$$

• Chebyshev's WLLN: 两两不相关, $var(X_i) \leq M, \forall i,$ 则

$$\frac{S_n - ES_n}{n} \stackrel{P}{\to} 0. \quad (5.1.12)$$

• Markov's WLLN: $var(S_n) = o(n^2)$, M

$$\frac{S_n - ES_n}{n} \stackrel{P}{\to} 0. \quad (5.1.12)$$

- $\chi \stackrel{ES_n}{=} \to a$, $M \stackrel{S_n}{=} \stackrel{P}{\to} a$.
- 习题五、26(4).

$$\xi_n \stackrel{P}{\to} \xi, \ \eta_n \stackrel{P}{\to} \eta \quad \Rightarrow \quad \xi_n + \eta_n \stackrel{P}{\to} \xi + \eta.$$

证明: $\forall \varepsilon > 0$,

$$P(\left|(\xi_n + \eta_n) - (\xi + \eta)\right| \ge \varepsilon)$$

$$\le P(\left|\xi_n - \xi\right| \ge \frac{\varepsilon}{2}) + P(\left|\eta_n - \eta\right| \ge \frac{\varepsilon}{2}).$$

• Bernoulli's WLLN (5.1.14), Poisson's WLLN (5.1.16).

习题五、44. $f:[0,1] \to \mathbb{R}$ 连续. 则存在多项式 f_n 使得 $f_n \stackrel{[0,1]}{\Rightarrow} f$.

LLN:
$$\frac{S_n}{n} \to x \Rightarrow f\left(\frac{S_n}{n}\right) \approx f(x)$$
(直观).

• $\mathfrak{P}f_n$:

$$f_n(x) = Ef\left(\frac{S_n}{n}\right) = \sum_{k=0}^n f\left(\frac{k}{n}\right) C_n^k x^k (1-x)^{n-k}.$$

• $\forall \varepsilon > 0, \exists \delta$ 使得

若
$$|y-z| < \delta$$
, 则 $|f(y)-f(z)| \leq \frac{1}{2}\varepsilon$.



•
$$\diamondsuit A_n = \left\{ \left| \frac{S_n}{n} - x \right| \geqslant \delta \right\} = \left\{ \left| S_n - nx \right| \geqslant n\delta \right\}.$$
 \mathbb{M}

$$f_n(x) - f(x) = E\left(f\left(\frac{S_n}{n}\right) - f(x)\right)\left(1_{A_n} + 1_{A_n^c}\right).$$

$$|f_n(x) - f(x)| \le CP(A_n) + \frac{1}{2}\varepsilon.$$

• $\sigma^2 = x(1-x) \leqslant \frac{1}{4}$: $\stackrel{\square}{=} n \geqslant N$ $\stackrel{\square}{=}$,

$$|f_n(x) - f(x)| \le C \frac{n\sigma^2}{n^2 \delta^2} + \frac{1}{2}\varepsilon$$

$$\le \frac{C}{4n\delta^2} + \frac{1}{2}\varepsilon \le \varepsilon, \quad \forall x \in [0, 1].$$

有界收敛定理: 假设 $\xi_n \stackrel{P}{\to} \xi$, 且 $P(|\xi_n| \leq M) = 1$, $\forall n$. 那么,

$$\lim_{n\to\infty} E\xi_n = E\xi.$$

• $i \exists A_n = \{ |\xi_n - \xi| > \varepsilon \}.$

$$\begin{split} E|\xi_n - \xi| = & E|\xi_n - \xi| \cdot 1_{A_n} + E|\xi_n - \xi| \cdot 1_{A_n^c} \\ \leqslant & 2M \cdot P(A_n) + \varepsilon. \end{split}$$

• $\xi_n \xrightarrow{1} \xi$:

$$\limsup_{n \to \infty} E|\xi_n - \xi| \leqslant \varepsilon, \quad \forall \varepsilon > 0.$$

 $\bullet |E\xi_n - E\xi| \le E|\xi_n - \xi| \to 0.$

定理5.3.1. 设 $X = X_1, X_2, \cdots$ i.i.d., $E|X| < \infty$, 则 $\frac{S_n}{n} \stackrel{P}{\to} EX$.

• $E|X| < \infty \Rightarrow \lim_{x \to \infty} xP(|X| > x) = 0.$

$$\frac{x}{2}P(|X|>x)\leqslant \int_{x/2}^x P(|X|>y)dy\stackrel{x\to\infty}{\longrightarrow} 0.$$

- $\frac{S_n}{n} EX \cdot 1_{\{|X| \leq n\}} \stackrel{P}{\to} 0.$ (见后面例*)
- $\bullet \ \mu_n = EX \cdot 1_{\{|X| \leqslant n\}} \to EX:$

$$E|X| \cdot \mathbf{1}_{\{|X| > n\}} = \int_0^\infty P(|X| \cdot \mathbf{1}_{\{|X| > n\}} > y) dy$$
$$= \int_0^\infty P(|X| > y \lor n) dy$$
$$= \int_0^n P(|X| > n) dy + \int_0^\infty P(|X| > y) dy \to 0.$$

例*. 假设 X_1, X_2, \cdots i.i.d., $\lim_{x\to\infty} xP(|X|>x)=0$, 则

$$\frac{S_n}{n} - EX \cdot 1_{\{|X| \le n\}} \stackrel{P}{\to} 0.$$

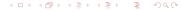
• 截断:

$$T_n = X_1 \cdot 1_{\{|X_1| \le n\}} + \dots + X_n \cdot 1_{\{|X_n| \le n\}}.$$

则,
$$P(S_n \neq T_n) \leq nP(X \neq X \cdot 1_{\{|X| \leq n\}}) = nP(|X| > n) \to 0.$$

• 故,

$$P\left(\left|\frac{S_n}{n} - \mu_n\right| \geqslant \varepsilon\right) \leqslant P(S_n \neq T_n) + P\left(\left|\frac{T_n}{n} - \mu_n\right| \geqslant \varepsilon\right).$$



• $\operatorname{var}(X \cdot 1_{\{|X| \le n\}}) \le EX^2 \cdot 1_{\{|X| \le n\}} = o(n)$:

$$EX^{2} \cdot 1_{\{|X| \leq n\}} = \int_{0}^{\infty} P\left(\frac{X^{2}}{1_{\{|X| \leq n\}}} > x\right) dx$$
$$= \int_{0}^{n^{2}} P\left(\sqrt{x} < |X| \leq n\right) dx$$
$$\leq \int_{0}^{n} P\left(y < |X|\right) \cdot 2y dy.$$

• $P(|\frac{T_n}{n} - \mu_n| \geqslant \varepsilon) \to 0$:

$$P\left(\left|\frac{T_n}{n} - \mu_n\right| \geqslant \varepsilon\right) \leqslant \frac{1}{\varepsilon^2} \frac{\operatorname{var}(T_n)}{n^2} = \frac{1}{\varepsilon^2} \frac{\operatorname{var}(X \cdot 1_{\{|X| \leqslant n\}})}{n} \to 0.$$

例*(习题一38, 习题四6). 假设有n 种券, 集齐时间为 S_n , 则

$$\frac{S_n}{n \ln n} \stackrel{P}{\to} 1.$$

• $S_n = X_1 + \cdots + X_n$:

$$X_k \sim G(p_k), \quad p_k = \frac{n - (k - 1)}{n}.$$

• 若 $X \sim G(p)$,则

$$EX = \frac{1}{p}, \quad \text{var}(X) = \frac{1-p}{p^2} \le \frac{1}{p^2}.$$

于是,

$$ES_n = n \sum_{\ell=1}^n \frac{1}{\ell} \approx n \ln n, \quad \operatorname{var}(S_n) \leqslant n^2 \sum_{\ell=1}^n \frac{1}{\ell^2} \leqslant Cn^2.$$

• $\frac{ES_n}{n \ln n} \to 1$, $var(S_n) \leqslant Cn^2$:

$$P\left(\left|\frac{S_n - ES_n}{n \ln n}\right| \ge \varepsilon\right) \le \frac{\operatorname{var}(S_n)}{\varepsilon^2 n^2 (\ln n)^2} \to 0.$$

• $\frac{S_n - ES_n}{n \ln n} \stackrel{P}{\to} 0$, ix

$$\frac{S_n}{n \ln n} \stackrel{P}{\to} 1.$$

• 集齐一半的时间: $T_n = X_1 + \cdots + X_{n/2}$.

$$ET_n \approx n \sum_{\ell=n/2}^n \frac{1}{\ell} \approx n \ln 2$$
, $var(T_n) \leqslant n^2 \sum_{\ell=n/2}^n \frac{1}{\ell^2} \leqslant \delta_n n^2$.

$$\text{ tx}, P\left(\left|\frac{T_n - ET_n}{n \ln 2}\right| > \varepsilon\right) \leqslant \frac{\delta_n n^2}{\varepsilon n^2 (\ln 2)^2} \to 0, \quad \frac{T_n}{n \ln 2} \xrightarrow{P} 1.$$



几乎必然收敛.

§5.2 之几乎必然收敛、§5.4 强大数定律

• 如果

$$P\left(\lim_{n\to\infty}\frac{\xi_n}{\xi_n}=\xi\right)=1.$$

那么, 称 ξ_n 几乎必然收敛到 ξ , 记为 $\xi_n \stackrel{\text{a.s.}}{\to} \xi$. (定义5.2.5)

- $A = \{\omega : \lim_{n \to \infty} \frac{\xi_n(\omega)}{\xi_n(\omega)} = \xi(\omega)\}.$
- $\lim_{n\to\infty} \xi_n(\omega) = \xi(\omega)$ 不成立:

$$\exists \varepsilon > 0, \ \forall N \geq 1, \ \exists n \geq N \$$
 使得 $|\xi_n(\omega) - \xi(\omega)| > \varepsilon$.

$$\exists k \geq 1, \ \forall N \geq 1, \ \exists n \geq N \$$
使得 $|\xi_n(\omega) - \xi(\omega)| > \frac{1}{k}$.



• $\diamondsuit A_{n,\varepsilon} = \{ |\xi_n - \xi| > \varepsilon \}, \$ $\$

$$\left\{ \lim_{n \to \infty} \xi_n = \xi \right\}^c = \bigcup_{\varepsilon > 0} \bigcap_{N \geqslant 1} \bigcup_{n \geqslant N} A_{n,\varepsilon}$$
$$= \bigcup_{k \geqslant 1} \bigcap_{N \geqslant 1} \bigcup_{n \geqslant N} A_{n,\frac{1}{k}}.$$

当ε \ 0 时,

$$A_{\varepsilon} := \bigcap_{N \geqslant 1} \bigcup_{n \geqslant N} A_{n,\varepsilon} ,$$

$$\star \star = \bigcup_{\varepsilon > 0} A_{\varepsilon} = \bigcup_{k \geqslant 1} A_{\frac{1}{k}} = \lim_{k \to \infty} A_{\frac{1}{k}}.$$

• 结论:

$$\xi_{n}\overset{\mathrm{a.s.}}{\longrightarrow}\xi\text{ iff }P\left(A_{\varepsilon}\right)=0,\forall\varepsilon>0\text{ iff }P\left(A_{\frac{1}{k}}\right)=0,\forall k\geqslant1.$$

• 对任意事件列 A_1, A_2, \cdots , 令

$${A_n \text{ i.o.}} = \limsup_{n \to \infty} A_n := \bigcap_{N \geqslant 1} \bigcup_{n \geqslant N} A_n.$$

- Borel-Cantelli引理(引理5.4.1): 令 $s = \sum_{n=1}^{\infty} P(A_n)$.
- (1) 若 $s < \infty$, 则 $P(A_n \text{ i.o.}) = 0$.
- (2) 若 $s = \infty$, 且 A_1, A_2, \cdots 相互独立, 则 $P(A_n \text{ i.o.}) = 1$.
 - $\bullet \bigcup_{n \geqslant N} A_n \searrow : (1 p \leqslant e^{-p},)$

$$P(A_n \text{ i.o.}) = \lim_{N \to \infty} P\left(\bigcup_{n \ge N} A_n\right).$$
$$\prod_{n \ge N} (1 - P(A_n)) \le e^{-\sum_{n \ge N} P(A_n)} = 0.$$

总结:

• $\xi_n \stackrel{\text{a.s.}}{\to} \xi$ 的定义及等价条件:

$$P\left(\left\{\omega: \lim_{n\to\infty} \xi_n(\omega) = \xi(\omega)\right\}\right) = 1. \quad (定义、应用、验证)$$

$$P\left(A_{n,\varepsilon} \text{ i.o.}\right) = 0, \quad \forall \varepsilon > 0; \quad (验证)$$

$$P\left(A_{n,\frac{1}{k}} \text{ i.o.}\right) = 0, \quad \forall k \geqslant 1;$$

$$P\left(\bigcup_{n\geqslant N} A_{n,\varepsilon}\right) \searrow 0(N\to\infty), \quad \forall \varepsilon > 0.$$

其中,
$$A_{n,\varepsilon} = \{ |\xi_n - \xi| > \varepsilon \}.$$

• $\xi_n \stackrel{\text{a.s.}}{\to} \xi$ 的充分条件:

$$\sum_{n=1}^{\infty} P(A_{n,\varepsilon}) < \infty, \quad \forall \varepsilon > 0.$$



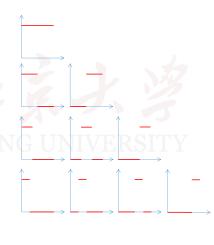
几乎必然收敛与其它收敛.

•
$$\xi_n \stackrel{\text{a.s.}}{\to} \xi \Rightarrow \xi_n \stackrel{P}{\to} \xi$$
. (定理5.4.1)

$$P\left(\bigcup_{n\geqslant N} A_{n,\varepsilon}\right) \geqslant P(A_{N,\varepsilon}),$$

$$A_{n,\varepsilon} = \{ |\xi_n - \xi| > \varepsilon \}.$$

- " $\xi_n \stackrel{P \vec{\boxtimes} r}{\longrightarrow} \xi \Rightarrow \xi_n \stackrel{\text{a.s.}}{\longrightarrow} \xi$ " 不成立. (例5.4.1)
- " $\xi_n \stackrel{\text{a.s.}}{\to} \xi \Rightarrow \xi_n \stackrel{r}{\to} \xi$ " 不成立. (例5.2.4)



- $\xi_n \xrightarrow{P} \xi$, 则存在子列 $\{n_k\}$ 使得 $\{n_k\}$ 使得 $\{n_k\}$ 。
- (1) ∃n₁ ≥ 1 使得

$$P(|\xi_n - \xi| > 1) < \frac{1}{2}, \quad \forall n \ge n_1;$$

(2) $\exists n_m > n_{m-1}$ 使得

$$P\left(|\xi_n - \xi| > \frac{1}{m}\right) < \frac{1}{2^m}, \quad \forall n \geqslant n_m.$$

(3) $\forall \varepsilon$, $\exists m$ 使得 $\frac{1}{m} < \varepsilon$. $\stackrel{.}{=} k \ge m$ 时, $\frac{1}{k} < \varepsilon$, $n_k \ge n_m$,

$$P(|\xi_{n_k} - \xi| > \varepsilon) \le P(|\xi_{n_k} - \xi| > \frac{1}{k}) < \frac{1}{2^k}.$$

(4) $\sum_{k} P(A_{k,\varepsilon}) < \infty \Rightarrow P(A_{k,\varepsilon}, \text{ i.o.}) = 0.$



- $\xi_n \xrightarrow{P} \xi$ iff $\forall \{n_k\}, \exists \{n_{k_i}\} \notin \exists \{n_{k_i}\}$ $\xrightarrow{\text{a.s.}} \xi$. \Longrightarrow : $\exists \text{$\mathbb{U}$}$
- ← : 反证法: 否则,

$$\exists \varepsilon, \delta > 0, \ n_1 < n_2 < \cdots$$
 使得 $P(|\xi_{n_k} - \xi| > \varepsilon) > \delta$.

- 对于 $\{\xi_{n_k}, k \geq 1\}$, 不存在收敛子列: 否则 $\xi_{n_{k_i}} \stackrel{\text{a.s.}}{\to} \xi$, 于是 $\xi_{n_{k_i}} \stackrel{P}{\to} \xi$. 与 $P\left(|\xi_{n_{k_i}} - \xi| > \varepsilon\right) > \delta$ 矛盾!
- 总结: $\xi_n \xrightarrow{\text{a.s.,r}} \eta \Rightarrow \xi_n \xrightarrow{P} \eta$, 反之不成立.

各种版本的SLLN.

强大数定律(Strong Law of Large Numbers, SLLN)形式:

 $若X_1, X_2, \cdots$ 满足***,则

$$\frac{S_n}{b_n} - a_n \stackrel{\text{a.s.}}{\to} 0. \quad (5.4.28)$$

定理 ((Borel-)Cantelli's SLLN, 定理5.4.2, 习题五、46)

若
$$X_1, X_2, \cdots$$
 相互独立, $E(X_i - EX_i)^4 \leq M$, $\forall i$, 则

$$\frac{S_n - ES_n}{n} \stackrel{\text{a.s.}}{\to} 0.$$

- 不妨设 $EX_i = 0$. $A_n = \{ |\frac{S_n}{n}| \ge \varepsilon \}$. 往证 $\sum_n P(A_n) < \infty$.
- 将证 $ES_n^4 \leq 3n^2M$, $\forall n$. 于是,

$$P(A_n) = P\left(S_n^4 \geqslant (n\varepsilon)^4\right) \leqslant \frac{1}{n^4\varepsilon^4} ES_n^4 \leqslant \frac{3M}{\varepsilon^4} \cdot \frac{1}{n^2}.$$



(1)
$$ES_n^4 = \sum_{i,j,k,\ell=1}^n EX_i X_j X_k X_\ell,$$

 $EX_r^4, EX_r^3 X_s, EX_r^2 X_s^2, EX_r^2 X_s X_t, EX_r X_s X_t X_u.$

$$(2) E^{\star} = 0, EX_r^4 \leqslant M,$$

$$EX_r^2X_s^2 = EX_r^2 \times EX_s^2 \leqslant \sqrt{EX_r^4}\sqrt{EX_s^4} \leqslant M.$$

(3) 因此,

$$ES_n^4 \leq {}_{n}M + C_n^2 C_1^2 M \leq 3n^2 M.$$



定理

若
$$X = X_1, X_2, \cdots$$
 独立同分布, $EX^2 < \infty$, 则

$$\frac{1}{n}S_n \stackrel{\text{a.s.}}{\to} EX.$$

• 不妨设EX = 0, $EX^2 = 1$. $\diamondsuit A_n = \{ \left| \frac{S_n}{n} \right| > \varepsilon \}$. 则

$$P(A_n) \leqslant \frac{1}{\varepsilon^2} E\left(\frac{S_n}{n}\right)^2 = \frac{1}{\varepsilon^2} \frac{nEX^2}{n^2} = \frac{1}{\varepsilon^2} \cdot \frac{1}{n}.$$

• $\frac{S_{m^2}}{m^2} \stackrel{\text{a.s.}}{\rightarrow} 0$:

$$\sum P(A_{m^2}) < \infty \Rightarrow P(A_{m^2} \text{ i.o.}) = 1.$$



•
$$|S_n| \le |S_{m^2}| + T_m$$
, $\forall m^2 \le n < (m+1)^2 = m^2 + 2m + 1$.

$$T_m := \max_{1 \le k \le 2m} |S_k^{(m)}|,$$

$$\not \pm \uparrow, \ S_k^{(m)} = X_{m^2+1} + \dots + X_{m^2+k}.$$

• $\frac{T_m}{m^2} \stackrel{\text{a.s.}}{\to} 0$. $\exists \mathbb{R} \in \mathbb{R}^n \xrightarrow{\text{a.s.}} 0$.

$$P\left(\left|\frac{S_{k}^{(m)}}{m^{2}}\right| > \varepsilon\right) \leqslant \frac{1}{\varepsilon^{2}} \frac{k}{m^{4}},$$

$$\Rightarrow P\left(\frac{|T_{m}|}{m^{2}} > \varepsilon\right) \leqslant 2m \cdot \frac{1}{\varepsilon^{2}} \frac{2m}{m^{4}} = \frac{4}{\varepsilon^{2}} \cdot \frac{1}{m^{2}}.$$

定理 (Kolmogorov's SLLN, 定理5.4.4)

假设
$$X = X_1, X_2, \cdots$$
 独立同分布, $E|X| < \infty$, 则

$$\frac{S_n}{n} \stackrel{\text{a.s.}}{\to} EX.$$

- 时间平均= 空间平均, (期望的含义).

- (1) 在 $A = \{\lim_n \frac{S_n}{n} \exists\}$ 上,

$$\frac{S_{n-1}}{n-1} - \frac{S_n}{n} = \frac{S_{n-1}}{(n-1)n} - \frac{X_n}{n}, \quad \text{ix } \frac{X_n}{n} \to 0.$$

(2) 习题五、45. 假设i.i.d. 则

$$\frac{X_n}{n} \stackrel{\text{a.s.}}{\to} 0 \quad \text{iff} \quad E|X| < \infty.$$

定理

假设
$$X = X_1, X_2, \cdots$$
 独立同分布, $EX = \infty$, 则

$$\frac{S_n}{n} \stackrel{\text{a.s.}}{\to} \infty.$$

- $X = X^{+} X^{-}$. 不妨设 $X \ge 0$.
- 给定M > 0, $\diamondsuit T_n = X_1 \wedge M + \cdots + X_n \wedge M$.

$$\frac{T_n}{n} \stackrel{\text{a.s.}}{\to} E(X \wedge M).$$

• $S_n \geqslant T_n$: $\forall M > 0$,

$$\liminf_{n \to \infty} \frac{S_n}{n} \geqslant E(X \land M), \quad \text{a.s.}$$



•
$$P(\Omega_M) = 1$$
. 令 $\hat{\Omega} = \bigcap_{M=1}^{\infty} \Omega_M$. 则 $P(\hat{\Omega}) = 1$, 且在 $\hat{\Omega}$ 上,

$$\liminf_{n\to\infty}\frac{S_n}{n}\geqslant E(X\wedge M),\quad \forall M.$$

• $E(X \wedge M) \rightarrow \infty$:

$$E(X \land M) = \int_0^\infty P(X \land M > x) dx$$
$$= \int_0^M P(X > x) dx \to \infty.$$

i.i.d. 序列总结:

● EX 有意义, 则

$$\frac{S_n}{n} \stackrel{\text{a.s.}}{\longrightarrow} EX.$$

• $EX^{\pm} = \infty$: 柯西分布(例3.3.5),

$$\frac{S_n}{n} \stackrel{d}{=} X_1.$$

SLLN的应用.

例5.3.1. 样本均值& 样本方差.

- 数据: *X*₁, *X*₂, · · · 独立同分布.
- 样本均值:

$$\bar{X} = \frac{1}{n} S_n \stackrel{\text{a.s.}}{\to} EX.$$

• 样本方差: $\frac{1}{n-1} \sum_{i=1}^{n} (X_i - \bar{X})^2 \approx$

$$\frac{1}{n} \sum_{i=1}^{n} (X_i - \bar{X})^2 = \frac{1}{n} \sum_{i=1}^{n} \frac{X_i^2}{X_i^2} - \bar{X}^2$$

$$\stackrel{\text{a.s.}}{\to} EX^2 - (EX)^2 = \text{var}(X).$$



例5.3.2. $\int_0^1 f(x)dx = Ef(U)$.

• SLLN:

$$\frac{1}{n} \sum_{i=1}^{n} f(U_i) \stackrel{\text{a.s.}}{\to} \int_{0}^{1} f(x) dx.$$

• 高维:

$$\int_0^1 \cdots \int_0^1 f(x_1, \cdots, x_n) d\vec{x} = Ef(U_1, \cdots, U_n).$$

习题五、50. 假设 $f, g: [0,1] \to \mathbb{R}, 0 \le f < Cg.$ 求

$$\lim_{n\to\infty}\int_0^1\cdots\int_0^1\frac{f(x_1)+\cdots+f(x_n)}{g(x_1)+\cdots+g(x_n)}dx_1\cdots dx_n.$$

$$W_n = \frac{S_n}{T_n}$$
. 再次强调: $EW_n \neq \frac{ES_n}{ET_n}$.

• SLLN:

$$W_n = \frac{S_n/n}{T_n/n} \stackrel{\text{a.s.}}{\longrightarrow} \frac{EX}{EY} = \frac{\int_0^1 f(x)dx}{\int_0^1 g(x)dx} = w.$$

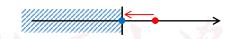
• 有界收敛定理:

$$EW_n \to Ew = w$$
.

§5.2 之依分布收敛, §5.3 之中心极限定理

- 权重的收敛.
- \emptyset 5.2.1: $\xi_n = \frac{1}{n}, \xi = 0$.

$$F_{\xi_n}(0) = 0, \quad F_{\xi}(0) = 1.$$



• 如果 $\forall x \in C(F_{\varepsilon})$ 都有

$$\lim_{n\to\infty} F_{\xi_n}(x) = F_{\xi}(x).$$

那么, 称 ξ_n 依分布收敛于 ξ , 记为 $\xi_n \stackrel{d}{\rightarrow} \xi$. (定义5.2.2 & 5.2.1)

• $\mathbb{R}\backslash C(F_{\xi})$ 可数.



依分布收敛与依概率收敛.

- $\xi_n \stackrel{P}{\to} \xi \Rightarrow \xi_n \stackrel{d}{\to} \xi$. (定理5.2.6)
- (1) $P(X \leqslant x) = P(X \leqslant x, Y \leqslant x) + P(X \leqslant x < Y)$: $|P(X \leqslant x) - P(Y \leqslant x)| \leqslant P(X \leqslant x < Y) + P(Y \leqslant x < X).$
- (2) 若 $|X Y| < \varepsilon$ 且 $X \le x < Y$, 则 $x < Y < x + \varepsilon$: $|P(X \le x) P(Y \le x)| \le P(|X Y| \ge \varepsilon) + P(|Y x| < \varepsilon).$
- (3) $|F_{\xi_n}(x) F_{\xi}(x)| \le P(|\xi_n \xi| \ge \varepsilon) + P(|\xi x| < \varepsilon).$



- " $\xi_n \xrightarrow{d} \xi \Rightarrow \xi_n \xrightarrow{P} \xi$ " 不成立. (例5.2.3)
- $\xi_n \xrightarrow{d} C \Rightarrow \xi_n \xrightarrow{P} C$. (定理5.2.7) 例, C = 0:

$$P(|\xi_n| < \varepsilon) \leqslant F_{\xi_n}(\varepsilon) - F_{\xi_n}(-\varepsilon) \to F_{\xi}(\varepsilon) - F_{\xi}(-\varepsilon) = 1.$$

• $\xi_n \stackrel{d}{\to} \xi \Rightarrow \hat{\xi}_n \stackrel{\text{a.s.}}{\to} \hat{\xi}$, 其中

$$\hat{\xi}_n := F_{\xi_n}^{-1}(U) \stackrel{d}{=} \xi_n, \quad \hat{\xi} := F_{\xi}^{-1}(U) \stackrel{d}{=} \xi.$$

(证明不要求)

• 有界收敛定理: ξ_n 有界, $\stackrel{d}{\rightarrow} \xi$, 则 $E\xi_n \rightarrow E\xi$.



依分布收敛的等价条件.

$$\xi_n \xrightarrow{d} \xi \text{ iff } Ef(\xi_n) \to Ef(\xi), \forall f \in \mathbb{F}.$$

- \mathbb{F}_1 : $f = 1_{(-\infty,b]}$, 其中 $b \in C(F_{\xi})$. (定义)
- \mathbb{F}'_1 : $f = 1_{(-\infty,b]}$, 其中 $b \in C(F_{\xi})$ 的某稠子集.
- \mathbb{F}_2 : $f = \mathbb{1}_{(a,b]}$, $\not = \mathbb{P}_1$, $f \in C(F_{\xi})$.
- F₂: 阶梯函数

$$f = \sum_{i=1}^{n} \lambda_i \cdot 1_{(a_i, b_i]},$$

其中, $a_i, b_i \in C(F_{\xi})$, $1 \leq i \leq n$.



- \mathbb{F}_3 : $f: \mathbb{R} \to \mathbb{R}$ 有界连续. \Rightarrow : (假设 \xrightarrow{d} , 定理5.2.2 & 5.2.3).
- (1) 取a 使得 $\pm a \in C(F_{\xi})$,

$$P(|\xi| \geqslant a) \leqslant \varepsilon \Rightarrow P(|\xi_n| \geqslant a) \leqslant 2\varepsilon, \ \forall n \geqslant N.$$

(2) 取阶梯函数g:

$$|f(x)-g(x)|\leqslant \varepsilon, \ \forall |x|\leqslant a; \qquad g(x)=0, \ |x|>a.$$

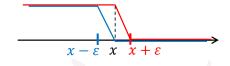
(3) $Ef(\xi_n) \cdot 1_{\{|\xi_n| \le a\}} - Ef(\xi) \cdot 1_{\{|\xi| \le a\}} = r_n + s_n + t_n$:

$$\begin{split} r_n = & \star - Eg(\xi_n) \cdot 1_{\{|\xi_n| \leqslant a\}}, \quad t_n = Eg(\xi) \cdot 1_{\{|\xi| \leqslant a\}} - \star, \\ s_n = & Eg(\xi_n) \cdot 1_{\{|\xi_n| \leqslant a\}} - Eg(\xi) \cdot 1_{\{|\xi| \leqslant a\}} = Eg(\xi_n) - Eg(\xi). \end{split}$$

(4) $|Ef(\xi_n) - Ef(\xi)| \leq 3M\varepsilon + \varepsilon + \varepsilon + |s_n|$.

• \mathbb{F}_3 : $f: \mathbb{R} \to \mathbb{R}$ 有界连续. \Leftarrow :

$$(1) g \leqslant 1_{(-\infty,x]} \leqslant f.$$



$$(2) \ 1_{(-\infty,x]} \leqslant f:$$

$$F_{\xi_n}(x) \leqslant Ef(\xi_n) \to Ef(\xi) \leqslant F_{\xi}(x+\varepsilon).$$

(3) $g \leq 1_{(-\infty,x]}$:

$$F_{\xi_n}(x) \geqslant Eg(\xi_n) \to Eg(\xi) \geqslant F_{\xi}(x - \varepsilon).$$

(4) $x \in C(F_{\xi})$:

$$\lim_{\varepsilon \searrow 0} F_{\xi}(x+\varepsilon) = \lim_{\varepsilon \searrow 0} F_{\xi}(x-\varepsilon) = F_{\xi}(x).$$

• \mathbb{F}_4 : f = 三角函数

$$\cos(tx), \quad \sin(tx), \quad t \in \mathbb{R}.$$

• 定理5.2.4 & 5.2.5 (证明不要求).

$$\xi_n \stackrel{d}{\to} \xi$$
 iff 特征函数收敛: $f_{\xi_n}(t) \to f_{\xi}(t)$, $\forall t$.

• 定理5.2.5, §4.5 连续性定理: 若

$$f_{\xi_n}(t) \to f(t), \ \forall t, \ \ \exists f \ \exists t = 0 \ \not\equiv \not\equiv.$$

则, f 是特征函数, ξ_n 依分布收敛. (证明不要求)

极限定理.

习题五、35. 当
$$\delta_n := \max_{1 \leq k \leq n} p_k^{(n)} \to 0, \ \lambda_n := \sum_{k=1}^n p_k^{(n)} \to \lambda$$
 时,

$$B(n, \vec{p})^{(n)} := B(1, p_1^{(n)}) * \cdots * B(1, p_n^{(n)}) \stackrel{d}{\to} P(\lambda).$$

• $B(n, \vec{p}^{(n)})$ 的特征函数:

$$f_n(t) = \prod_{k=1}^n \left((1 - p_k^{(n)}) + p_k^{(n)} e^{it} \right)$$

$$= \prod_{k=1}^n \left(1 + p_k^{(n)} z \right), \quad \sharp r = e^{it} - 1.$$

• $P(\lambda_n) \to P(\lambda)$: 令 $g_n(t)$ 表示 $P(\lambda_n)$ 的特征函数,则

$$g_n(t) = \prod_{k=1}^n e^{p_k^{(n)}z} = e^{\lambda_n z} \to e^{\lambda z}.$$



$$\bullet \ a_k = 1 + p_k^{(n)} z, \ b_k = e^{p_k^{(n)} z},$$

$$\begin{aligned} \|\mathbf{a}_{k}\| &= \|1 + p_{k}^{(n)}z\| = \|(1 - p_{k}^{(n)}) \cdot 1 + p_{k}^{(n)} \cdot e^{it}\| \leq 1. \\ \|b_{k}\| &= \|e^{p_{k}^{(n)}z}\| = \|e^{p_{k}^{(n)}(\cos t - 1) + ip_{k}^{(n)}\sin t}\| = e^{p_{k}^{(n)}(\cos t - 1)} \leq 1. \end{aligned}$$

• $f_n(t) = \prod_{k=1}^n \frac{a_k}{a_k}, g_n(t) = \prod_{k=1}^n \frac{b_k}{b_k},$ by

$$||f_n(t) - g_n(t)|| \le \sum_{k=1}^n ||a_k - b_k||.$$
 (5.5.13)

• 归纳法:

$$\begin{split} & \| \prod_{k=1}^{n} a_k - \prod_{k=1}^{n-1} a_k \times b_n \| \leqslant 1 \times \| a_n - b_n \|; \\ & \| \prod_{k=1}^{n-1} a_k \times b_n - \prod_{k=1}^{n-1} b_k \times b_n \| \leqslant \| \prod_{k=1}^{n-1} a_k - \prod_{k=1}^{n-1} b_k \| \times 1. \end{split}$$

• $i\exists s = p_k^{(n)} z$, $\emptyset |a_k = 1 + s$, $b_k = e^s$, $i \otimes |a_k - b_k| \le ||s||^2$:

$$||a_k - b_k|| = ||\sum_{\ell=2}^{\infty} \frac{s^{\ell}}{\ell!}|| \leqslant \sum_{\ell=2}^{\infty} \frac{||s||^{\ell}}{\ell!} \leqslant \sum_{\ell=2}^{\infty} \frac{||s||^2}{\ell!}$$
$$\leqslant ||s||^2 (e - 2) \leqslant (p_k^{(n)})^2 ||z||^2.$$

- 注: 以上的估计并不精细.
- 故,

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$$||f_n(t) - g_n(t)|| \le \sum_{k=1}^n \delta_n p_k^{(n)} ||z||^2 \le 4\delta_n \lambda_n \to 0.$$



定理5.3.1. 设 $X=X_1,X_2,\cdots$ i.i.d., $E|X|<\infty$, 则 $\frac{S_n}{n}\stackrel{P}{\to}EX$. 证明 方法II

• 不妨设EX = 0. 只需证

$$f_n(t) = f_{\frac{S_n}{n}}(t) \to 1 = e^{it \times 0}.$$

• 记 $f = f_X$. 于是

$$f_n(t) = E \exp\left\{it\frac{X_1 + \dots + X_n}{n}\right\} = \left(Ee^{i\frac{t}{n}X}\right)^n = f\left(\frac{t}{n}\right)^n.$$

• 注: 根据有界收敛定理, $Ee^{i\frac{t}{n}X} \rightarrow e^{i\times 0} = 1$. 往估计收敛速度.



- $e^{is} = 1 + is + \delta(s)$, $\|\delta(s)\| \le \min\{2|s|, \frac{1}{2}s^2\}$. (引理5.5.1)
- $f(\frac{t}{n})$:

$$E\left(1+i\frac{t}{n}X+\delta\left(\frac{t}{n}X\right)\right)=1+0+\underline{E}\delta\left(\frac{t}{n}X\right).$$

• $||E\delta(\frac{t}{n}X)|| = o(\frac{1}{n})$:

$$\begin{split} \left\| E\delta\left(\frac{t}{n}X\right) \right\| = & E\left\| \delta\left(\frac{t}{n}X\right) \right\| \cdot \mathbf{1}_{\{|X| > M\}} + E\left\| \delta\left(\frac{t}{n}X\right) \right\| \cdot \mathbf{1}_{\{|X| \leqslant M\}} \\ \leqslant & \frac{2t}{n}E|X|\mathbf{1}_{\{|X| > M\}} + \frac{t^2}{2n^2}M^2. \end{split}$$

- $f(\frac{t}{n})^n = (1 + o(\frac{1}{n}))^n \to 1.$
- $\bullet \ \ \frac{1}{n}S_n \stackrel{d,P}{\to} 0.$

定理 (Lindeberg-Levy 中心极限定理, 定理5.4.4)

假设 X_1, X_2, \cdots 独立同分布. $0 < var(X) < \infty$. 则

$$S_n^* \stackrel{d}{\to} Z \sim N(0,1).$$

• 不妨设EX = 0, $EX^2 = 1$. $idf = f_{X_1}$, 则

$$f_{S_n^*}(t) = E \exp\left\{it\frac{S_n}{\sqrt{n}}\right\} = f\left(\frac{t}{\sqrt{n}}\right)^n.$$

• (记 $s = \frac{t}{\sqrt{n}}$.) 往证

$$f(s) = Ee^{isX} = 1 - \frac{1}{2}s^2 + o(s^2).$$
于是 $f_{S_n^*}(t) = (1 - \frac{t^2}{2n} + o(\frac{1}{n}))^n \to e^{-t^2/2}.$

引理5.5.1. (记y = sX.)

$$|e^{iy} - 1| \le |y|, |e^{iy} - 1 - iy| \le \frac{|y|^2}{2} (5.5.10),$$

 $|e^{iy} - 1 - iy + \frac{y^2}{2}| \le \frac{|y|^3}{6} (5.5.11).$

• $e^{iy} = 1 + iy + \frac{1}{2}(iy)^2 + \varphi(y)$.

$$\begin{split} \varphi(y) = & e^{iy} - 1 - iy - \frac{1}{2}(iy)^2 \\ = & \int_0^y i \underline{e}^{iz} dz - \int_0^y i \cdot \underline{1} dz - \frac{1}{2}(iy)^2 \\ = & \int_0^y i \underline{\int_0^z i e^{iw} dw} dz - \int_0^y \int_0^z i^2 dw dz \\ = & \int_0^y \int_0^z i^2 (e^{iw} - 1) dw dz = \int_0^y \int_0^z \int_0^w i^3 e^{iu} du dw dz. \end{split}$$

• $\|\varphi(y)\| \le \frac{y^2}{6}$; $\|\varphi(y)\| \le \frac{|y|^3}{6}$:

$$\varphi(y) = \int_0^y \int_0^z i^2 (e^{iw} - 1) dw dz = \int_0^y \int_0^z \int_0^w i^3 e^{iu} du dw dz.$$

• $\mathbb{R}y = sX$. \mathbb{N}

$$\begin{split} e^{isX} = &1 + isX + \frac{1}{2}(isX)^2 + \varphi(sX), \\ f(s) = &Ee^{isX} = &1 + 0 - \frac{1}{2}s^2 + E\varphi(sX). \end{split}$$

• 由 $\|\varphi(y)\| \le y^2$, $\|\varphi(y)\| \le \frac{|y|^3}{6}$ 可推出 $\|E\varphi(sX)\| = o(s^2)$:

$$\begin{split} \|E\varphi(sX)\| \leqslant & E\|\varphi(sX)\| \cdot 1_{\{|X|>M\}} + E\|\varphi(sX)\| \cdot 1_{\{|X|\leqslant M\}} \\ \leqslant & E(sX)^2 \cdot 1_{\{|X|>M\}} + E\frac{|sX|^3}{6} \cdot 1_{\{|X|\leqslant M\}} \\ \leqslant & s^2 EX^2 \cdot 1_{\{|X|>M\}} + |s|^3 \cdot \frac{M^3}{6}. \end{split}$$

• 注:

$$f(s) = 1 - \frac{1}{2}s^2 + E\varphi(sX) = 1 - \frac{1}{2}s^2 + o(s^2).$$

• 固定t. 取 $s = \frac{t}{\sqrt{n}}$, $\diamondsuit n \to \infty$. 则

$$f(\frac{t}{\sqrt{n}}) = 1 - \frac{t^2}{2n} + \varepsilon_n,$$

其中, $\varepsilon_n := E\varphi\left(\frac{t}{\sqrt{n}}X\right) = o(\frac{t}{\sqrt{n}}) = o(\frac{1}{n}).$

• $S_n^* \stackrel{d}{\rightarrow} Z \sim N(0,1)$:

$$f_{S_n^*}(t) = f\left(\frac{t}{\sqrt{n}}\right)^n = \left(1 - \frac{t^2}{2n} + \varepsilon_n\right)^n \to e^{-\frac{t^2}{2}}.$$



• 中心极限定理(Central Limit Theorem, CLT)形式:

若
$$X_1, X_2, \cdots$$
 满足***,则

$$S_n^* \xrightarrow{d} Z \sim N(0,1).$$
 (5.1.11)

- Lindeberg-Levy 版本: *** = "i.i.d., $0 < var(X) < \infty$ ".
- $\frac{1}{\sqrt{n}}(S_n ES_n) \stackrel{d}{\to} \sigma Z \sim N(0, \sigma^2).$
- $\vec{x}\vec{X} = \vec{X}_1, \vec{X}_2, \cdots$ i.i.d., $\vec{\mu} = E\vec{X}, \Sigma = (\text{cov}(X_i, X_j))$. \mathbb{N}

$$\frac{1}{\sqrt{n}}(\vec{S}_n - n\vec{\mu}) \stackrel{d}{\to} \sqrt{\Sigma}\vec{Z} \sim N(\vec{0}, \Sigma).$$

• Berry-Esseen's bound: 假设 $E|X|^3 < \infty$. 那么,

$$|F_{S_n^*}(x) - \Phi(x)| \leqslant \frac{3E|X^*|^3}{\sqrt{n}}, \quad \forall x.$$
 (了解即可.)

CLT的应用.

例: 经验分布函数.

SLLN:
$$F_n(x) = \frac{1}{n} \sum_{i=1}^n 1_{\{X_i \le x\}} \stackrel{\text{a.s.}}{\to} F(x).$$

• CLT:

$$\sqrt{n}(F_n(x) - F(x)) \xrightarrow{d} \sqrt{F(x)(1 - F(x))}Z$$

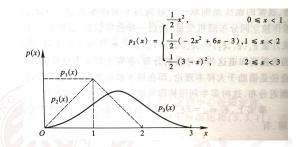
$$\sim N(0, F(x)(1 - F(x))).$$

• (了解即可) $D_n = \max_x |F_n(x) - F(x)|$:

$$\sqrt{n}D_n \stackrel{d}{\to} \xi, \quad F_{\xi}(x) = \sum_{k=-\infty}^{\infty} (-1)^k e^{-2k^2 x^2}, \ \forall x > 0.$$



• \emptyset 5.3.4. X_1, X_2, \cdots i.i.d., $\sim U(0, 1)$.



• βS_n^* "视为" ~ N(0,1):

$$P(S_n \le x) = P(S_n^* \le x^*) \approx \Phi(x^*) = p, \ x^* = \frac{x - ES_n}{\sqrt{\text{var}(S_n)}}.$$

n, x, p 满足关系式:

$$P(S_n \leqslant x) = p.$$



例5.3.5. 在近似计算中保留5位小数. 考虑 10^4 个数, 近似值之和与真值之和的误差不超过 10^{-3} 的概率.

- 近似值与真值的误差 $X \sim U(-0.5 \times 10^{-5}, 0.5 \times 10^{-5})$.
- $n = 10^4$, $x = 10^{-3}$, $\Re p$.
- 目标:

$$P(|S_n| \le 10^{-3}) = P(|S_n^*| \le x^*) = p,$$

其中,
$$x^* = \frac{10^{-3}}{\sqrt{n \cdot \text{var}(X)}} = \frac{10^{-3}}{\sqrt{10^4 \cdot 10^{-5}} \sqrt{\frac{1}{12}}} = 2\sqrt{3} = 3.46 \cdot \cdot \cdot \cdot$$

• 查表: $\Phi(3.46) = 99.97299\%$, 故 $p \ge 2\Phi(3.46) - 1 = 99.94598\%$.



例2.4.4. 人寿保险. 死亡率p = 0.005, 参保数 $n = 10^4$. 估算:

$$P($$
死亡人数 = 40), $P($ 死亡人数 \leq 70).

- $P(A) = C_{10000}^{40} p^{40} q^{9960} = \underline{0.0214}.$
- $x^* = \frac{x-np}{\sqrt{npq}}$:

$$P(A) \approx \varphi(40^*) \frac{1}{\sqrt{npq}} = \underline{0.0207}.$$

• 连续修正: $P(A) = P(39.5 \le S_n \le 40.5)$, $x_1 = 39.5^* \approx -1.49$, $x_2 = 40.5^* \approx -1.35$:

$$P(A) = P(39.5 \le S_n \le 40.5) = P(x_1 \le S_n^* \le x_2)$$

 $\approx 0.93189 - 0.91149 = 0.0204.$

- 连续修正: $P(B) = P((-0.5)^* \le S_n^* \le (70.5)^*) \approx 0.998193.$
- 如果不连续修正: $P(0^* \le S_n \le 70^*) \approx 0.997744$.

例2.4.6. 车间有200台车床, 每台的功率为1 千瓦, 开动频率为60%. 要求正常生产的把握至少为99.9%, 问: 需多少千瓦电力?

- $\Box \exists n = 200, p = 99.9\%, \ \exists x.$
- P(X = 1) = 1 P(X = 0) = 0.6. n = 200. p = 99.9%.
- $\exists \vec{\pi} : x^* = \frac{x 200 * 0.6}{\sqrt{200 * 0.6 * 0.4}},$

$$P(S_n \le x) = P(S_n^* \le x^*) \ge 99.9\% = p.$$

- $ilde{\Phi}$ 8: Φ (3.09) = 99.8999%, Φ (3.1) = 99.90324%.
- $\star \geq 3.1$, $x \geq 120 + 3.1\sqrt{48} \in (141, 142)$. $\equiv 142 \neq \mathbb{Z}$.



例5.1.2. 往常的市场占有率为15%. 为保证调查结果与真实值的误差不超过0.01 的概率至少为95%. 至少需调查多少人?

- P(X = 1) = 1 P(X = 0) = q = 0.15. 目标:

$$P\left(\left|\frac{S_n}{n} - q\right| \le 0.01\right) \ge 0.95.$$

$$P\left(\left|S_n^*\right| \le \frac{0.01\sqrt{n}}{\sqrt{q(1-q)}} = x^*\right) \ge 0.95.$$

- $\Phi(x^*) \ge \frac{1}{2}(1 + 0.95) = 0.975$, $\hat{\Xi}$ $\xi = 0.975$, $\hat{\Xi}$
- $n \ge 196^2 q(1-q)$, 至少调查4899 人.



非i.i.d.情形的CLT.

§5.5 中心极限定理

• 研究对象:

$$S_n^* = \sum_{k=1}^n \frac{X_k - \mu}{\sqrt{\operatorname{var}(S_n)}} = \sum_{k=1}^n Y_k \overset{\text{iid}}{=} \overset{\text{lif}}{=} \sum_{k=1}^n \frac{1}{\sqrt{n}} X_k^*.$$

• 假设 X_1, X_2, \cdots 相互独立, $EX_k = \mu_k, \operatorname{var}(X_k) = \sigma_k^2$. 记

$$B_n^2 = \sum_{k=1}^n \sigma_k^2.$$

• 目标:

$$S_n^* = \frac{S_n - \sum_{k=1}^n \mu_k}{B_n} = \sum_{k=1}^n \frac{X_k - \mu_k}{B_n} \xrightarrow{d} Z \sim N(0, 1).$$

$$S_n^* = \sum_{k=1}^n Y_k, \quad \sharp \Psi, Y_k := \frac{X_k - \mu_k}{B_n}.$$

• Lindeberg 条件 (5.5.2):

$$\sum_{k=1}^{n} E \frac{\mathbf{Y}_{k}^{2} \cdot 1_{\{|\mathbf{Y}_{k}| > \varepsilon\}}}{1_{\{|\mathbf{Y}_{k}| > \varepsilon\}}} \to 0, \ \forall \varepsilon > 0.$$

- Lindeberg $\Re H \Longrightarrow P(\max_{1 \le k \le n} Y_k > \varepsilon) \to 0 \quad (n \to \infty).$
- Feller $\Re + (5.5.3)$: $\frac{\max_{1 \le k \le n} \sigma_k}{B_n} \to 0$.

$$\frac{\max_{1 \le k \le n} \sigma_k}{B_n} \to 0$$
 iff $B_n \to \infty$ 且 $\frac{\sigma_n}{B_n} \to 0$. (定理5.5.1)

• Lindeberg-Feller CLT (定理5.5.2):

(5.5.2) iff
$$S_n^* \stackrel{d}{\to} Z \sim N(0,1)$$
 \coprod (5.5.3).