

1. Classical Limit Theory

① Jacob Bernoulli : LLN $EX < \infty \Rightarrow \bar{X} \xrightarrow{\text{a.s.}} \mu$

② de Moire & Lindenberg : CLT $\begin{aligned} EX < \infty \\ \text{Var } X < \infty \end{aligned} \Rightarrow \frac{S_n - n\mu}{\sigma/\sqrt{n}} \xrightarrow{\text{d}} N(0, 1)$

③ Khinchin & Kolmogorov : law of iterated logarithm

$$\limsup_{n \rightarrow \infty} \frac{S_n - n\mu}{\sqrt{2n \log \log n}} = 1 \text{ a.s.}$$

$$\liminf_{n \rightarrow \infty} \frac{S_n - n\mu}{\sqrt{2n \log \log n}} = -1 \text{ a.s.}$$

$$\arg \limsup(\cdot) = \arg(-\liminf)(\cdot)$$

2. Central Limit Theorem & Related Problems

assume $\mu = 0$, observe CLT $\Leftrightarrow \sup_x |P\left(\frac{S_n}{\sigma} \geq x\right) - (1 - \Phi(x))| \rightarrow 0$

↑
usually unknown
↓
estimate

① Berry Esseen Inequality: abt the abs error of the estimation

$$|A - B| \leq \frac{0.4748}{\sqrt{n}} \frac{E|X|^3}{\sigma^3}$$

also called Kolmogorov distance.

② Cramér Moderate Deviation Theorem: Relative error.

$$A/B = \exp\left(x^2 \lambda / (\sqrt{n})\right) \left|1 + O\left(\frac{|x|}{\sqrt{n}}\right)\right| \quad \text{if } Ee^{t_0|x|} < \infty \text{ for } t_0 > 0$$

where $\lambda(t)$ is Cramér's series. $x \geq 0$ and $x = o(n^{1/2})$

$$o(n^{1/6})$$

Yuri Linnik relaxed to $E(e^{t_0|x|}) < \infty$ afterwards

\star Depend on how many moments of X can match those of standard norm.

(3) Comparison b/w Abs err & Rel err

$$P\text{-Value} = P(T \geq x_n)$$

$$T \xrightarrow{d} N(0, 1)$$

$$\hat{P} = 1 - \Phi(x_n)$$

$$\hat{P} - P \rightarrow 0$$

~~\hat{P}~~ \uparrow
unknown, but always very small
 \downarrow

$$P / \hat{P} \rightarrow 1$$

References:

{ Petrov. (1972). Sum of Ind. R.V.
Petrov. (1995) Limit Theorems of Prob. Theory

Singularity prob. of Random Bernoulli matrices

$$A_n = (\varepsilon_{ij})_{n \times n}$$

Textbook: "Self-Normalized Process"

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Normal Approx. by Stein's Method

$$S_n = \sum_{i=1}^n X_i, \quad X_i \text{ indep. } E[X_i] = 0$$

$$P\left(\max_{1 \leq k \leq n} |S_k| \geq x\right) \leq E[S_n^2]/x^2 \quad (\text{maximum inequality})$$

Chapter 1 Inequalities

1. Chebychev's Inequalities

$$(a) \forall x > 0, \quad P(X \geq x) \leq E[X]/x$$

$$[\text{Pf. } 1(X \geq x) \leq X^+ / x]$$

$$P(X \geq x) \leq \frac{E(X 1(X \geq x))}{x}$$

$$(b) P(X \geq x) = P(e^{tX} \geq e^{tx}) \leq E(e^{tX})/e^{tx}$$

↑
transformation
using increasing non-neg. fn

↑ useful for
exponential ineq. for rv.

$$(c) P((x,y) \in A) = P(X \in A^1, Y \in A^2)$$

$$= E(1(X \in A^1, Y \in A^2))$$

$$\leq E\left(e^{sX+tY - \inf_{(x,y) \in A} (sx+ty)}\right)$$

when $(x,y) \in A$, $\exp(\cdot)$ should ≥ 1

2. Lyapunov Inequalities

Recall Hölder's Inequality:

$$E|XY| \leq (E|x|^p)^{\frac{1}{p}} (E|x|^q)^{\frac{1}{q}}, \quad E|x| \leq (E|x|^p)^{\frac{1}{p}}$$

where $p > 1$, $\frac{1}{p} + \frac{1}{q} = 1$

use higher moment for a upper bound

$$0 \leq r \leq s \leq t$$

$$E|x|^s \leq (E|x|^r)^{\frac{t-s}{t-r}} (E|x|^t)^{\frac{s-r}{t-r}}$$

$$r=1, s=2, t=4$$

$$E(X^2) \leq (E|x|)^{\frac{2}{3}} (E|X^4|)^{\frac{1}{3}} \Rightarrow (E|x|)^{\frac{2}{3}} \geq \frac{E(X^2)}{(E|X^4|)^{\frac{1}{3}}}$$

↑
gives a lower bound using higher moments

Proof: (Make Use of Hölder's Inequality)

$$\text{Let } Y = |x|^p, Z = |x|^r$$

$$\text{then } E(|x|^r) \leq E(|x|^p)^{\frac{r}{p}}$$

since $p \geq 1$, we have $rp \geq r$, let $rp = s$,

$$\text{then } E(|x|^r) \leq E(|x|^s)^{\frac{r}{s}} \quad \text{i.e. } E(|x|^r)^{\frac{1}{r}} \leq E(|x|^s)^{\frac{1}{s}}$$

$$0 < r < s < t$$

3. Kinball's Inequality

(a) $g(x) \uparrow, h(x) \uparrow \quad E(g(x)h(x)) \geq E(g(x))E(h(x))$ 单调性相同

[Introduce $Y \stackrel{d}{=} X$ and $Y \perp\!\!\!\perp X$

then $E(g(x))E(h(x)) = E(g(x))E(h(Y)) = \frac{1}{2} (E(g(x)h(Y)) + E(g(Y)h(x)))$

$$\begin{aligned} & 2E(g(x))E(h(x)) \\ &= E(g(x)h(Y)) + E(g(Y)h(x)) \\ &\vdash \leq E(g(x)h(X) + g(Y)h(Y)) \end{aligned}$$

$$\begin{aligned} & g(x)h(Y) + g(Y)h(X) - (g(x)h(x) + g(Y)h(Y)) \\ &= (g(x) - g(Y))(h(x) - h(Y)) \leq 0 \quad \text{#} \end{aligned}$$

(b) $g(x) \uparrow \quad h(x) \downarrow \quad E(g(x)h(x)) \leq E(g(x))E(h(x))$ 单调性相反

4. Bernstein-Hoeffding's Inequality

$$E(X_i) \leq 0, \quad X_i \leq a, \quad a \geq 0. \quad \sum_{i=1}^n E X_i^2 \leq B_n^2$$

then

$$\textcircled{1} \quad P(S_n \geq x) \leq \exp\left(-\frac{x^2}{2B_n^2 + ax}\right)$$

$$\textcircled{2} \quad P(S_n \geq x) \leq \exp\left(-\frac{B_n^2}{ax^2}\right) \left[\left(1 + \frac{ax}{B_n^2}\right) \ln\left(1 + \frac{ax}{B_n^2}\right) - \frac{ax}{B_n^2} \right]$$

$$\underbrace{(1+y) \ln(1+y) - y}_{\downarrow}$$

$$0 + \frac{1}{1+y} y - \frac{1}{(1+y)^2} y^2/2 + \frac{1}{(1+y)^3} y^3/6 \dots$$

$$\begin{aligned}
 & \text{when } y \rightarrow 0 \\
 &= y - \frac{y^2}{2} \left(\frac{1}{1+y} \right)^2 + \frac{y^3}{6} \left(\frac{1}{1+y} \right)^3 - \dots - y \\
 &\geq \frac{y^2}{2(1+y)}
 \end{aligned}$$

$$\forall t > 0, P(S_n \geq \lambda) \leq e^{-t\lambda} E e^{tS_n}$$

$$E e^{tS_n} = \prod_{i=1}^n E e^{tX_i}$$

$$e^x \leq 1 + \lambda + \frac{\lambda^2}{2} C$$