

(*) ■ X_1, X_2, \dots iid $E e^{t_0 X_1} < \infty, t_0 > 0$

then $\forall x > E(X_1)$

$$P\left(\frac{S_n}{V_n} \geq x\right) \xrightarrow{t \rightarrow 0} \inf_{t \geq 0} e^{-tx} E e^{tX_1}$$

$$T_n \Leftrightarrow \frac{S_n}{V_n}$$

■ $V_n^2 = \sum_i^n X_i^2, X_i > \frac{EX_1}{\sqrt{EX_1^2}}$ if $EX_1^2 = \infty$ then RHS = 0
(and/or $EX_1 = \infty$)

$$P\left(\frac{S_n}{V_n \sqrt{n}} \geq x\right)^{\frac{1}{n}} \rightarrow ?$$

↖ control order of denom. to be $O(n)$

■ $P\left(\frac{S_n}{V_n^2} \geq x\right) = P(S_n - xV_n^2 \geq 0) \quad (x > 0)$
no stat. meaning $= P\left(\sum_{i=1}^n (X_i - xX_i^2)\right)$ ↘ assume

$$X_i - xX_i^2 \leq C_x < \infty$$

Y_i has upper bound, therefore \exists mgf.
 $< \infty$

$$P\left(\sum (X_i - xX_i^2) \geq 0\right)^{\frac{1}{n}} \xrightarrow{(*)} \inf_{t \geq 0} E e^{t(X_1 - xX_1^2)}$$
 true for r.v. X

$$\forall \alpha > E(X_1 - xX_1^2), x > \frac{E(X_1)}{E(X_1^2)}, x > 0$$

Recall Cauchy Inequality: $xy \leq \frac{1}{2}(x^2 + y^2)$

$$xy = x\sqrt{c} \cdot \frac{y}{\sqrt{c}} \leq \frac{1}{2}(x^2 c + y^2/c)$$

$$xy = \frac{1}{2} \inf_{c>0} \left(\frac{cx^2 + y^2}{c} \right)$$

$$\text{write } x\sqrt{n} \cdot V_n = \frac{1}{2} \inf_{c>0} \left(\frac{cV_n^2 + x^2 n}{c} \right)$$

$$P(S_n \geq x V_n \sqrt{n}) = P(S_n \geq \frac{1}{2} \inf_{c>0} \left(\frac{c^2 V_n^2 + x^2 n}{c} \right))$$

$$= P \left(\bigcup_{c>0} \{ 2c S_n \geq c^2 V_n^2 + x^2 n \} \right)$$

$$= P \left(\bigcup_{c>0} \left\{ \sum_{i=1}^n 2c X_i - c^2 X_i^2 \geq x^2 n \right\} \right)$$

$$P(S_n \geq x V_n \sqrt{n})^{\frac{1}{n}} \geq \sup_{c>0} P \left(\frac{\sum_i 2c X_i - c^2 X_i^2}{n} \geq x^2 \right)^{\frac{1}{n}}$$

$$\text{by (*)} \rightarrow \sup_{c>0} \inf_{t \geq 0} e^{-t x^2} E[e^{t(\sum_i 2c X_i - c^2 X_i^2)}]$$

$$a_{n,i} \geq c_i, n \rightarrow \infty$$

$$a_{n,i} \xrightarrow{n \rightarrow \infty} \max c_i$$

upper bound is also



(Shao, 1977) If $E\bar{X}=0$, $E\bar{X}_i^2 \mathbb{1}(|\bar{X}_i| \leq z)$ slowly varying

then $\forall \bar{X}_n \rightarrow \infty$, $\bar{Z}_n = O(\sqrt{n})$

$$\ln P\left(\frac{\bar{S}_n}{\sqrt{n}} \geq \bar{x}_n\right) \sim -\frac{\bar{x}_n^2}{2} \quad \text{i.e. } \forall 0 < \varepsilon < 1$$

$$e^{-(1+\varepsilon)\bar{x}_n^2/2} \leq P\left(\frac{\bar{S}_n}{\sqrt{n}} \geq \bar{x}_n\right) \leq e^{-\frac{(1-\varepsilon)\bar{x}_n^2}{2}}$$

← more robust
becuz no assumption
on moment existence

$$P\left(\frac{\bar{S}_n}{\sigma\sqrt{n}} \geq \bar{x}_n\right) \rightarrow 0 \iff Ee^{\bar{x}_n^2} < \infty \quad \star$$

■ $S_n = \sum X_i \mathbb{1}(|X_i| \leq \bar{z}_n) + \sum X_i \mathbb{1}(|X_i| > \bar{z}_n)$

$$\triangleq \sum X_{i,1} + \sum X_{i,2}$$

$$P\left(\frac{\bar{S}_n}{\sqrt{n}} \geq \bar{x}_n\right)$$

$$P(X+Y \geq x) \leq P(X \geq (1-\varepsilon)x) + P(Y \geq \varepsilon x)$$

$$\leq P\left(\frac{\sum X_{i,1}}{\sqrt{n}} \geq (1-\varepsilon)\bar{x}_n\right) + P\left(\frac{\sum X_{i,2}}{\sqrt{n}} \geq \varepsilon \bar{x}_n\right)$$

(2) (1)

① $P\left(\frac{\sum X_{i,2}}{\sqrt{n}} \geq \varepsilon \bar{x}_n\right) = P\left(\frac{\sum X_i \mathbb{1}(|X_i| > \bar{z}_n)}{\sqrt{n}} \geq \varepsilon \bar{x}_n\right)$

$$\leq P\left(\left(\sum \mathbb{1}(|X_i| > \bar{z}_n)\right)^{\frac{1}{2}} \geq (\varepsilon \bar{x}_n)^{\frac{1}{2}}\right)$$

$\sqrt{\sum a_i b_i} \leq \sqrt{\sum a_i^2} \sqrt{\sum b_i^2}$

ε_i indep. $P(\varepsilon_i = 1) = p_i$ $P(\varepsilon_i = 0) = 1-p_i$

$$P\left(\sum \varepsilon_i \geq x\right) \leq \left(\frac{3 \sum_{i=1}^n p_i}{x}\right)^x$$

$$\leq \left(\frac{3n P(|X_i| > \bar{z}_n)}{\varepsilon^2 \bar{x}_n^2}\right) \varepsilon^2 \bar{x}_n^2$$

$$\leq e^{-2\bar{x}_n^2} \quad \text{① } \frac{n}{\bar{x}_n^2 \bar{z}_n^2} \leq C$$

Here we first assume Second moment exists $EX_i^2 < \infty$

$$P(|X_i| > z_n) \leq \frac{EX_i^2 1(|X_i| > z_n)}{z_n^2}$$

$$P(|X_i| > z_n) = o\left(\frac{1}{z_n}\right)$$

Then when $\lambda(x) = EX_i^2 1(|X_i| \leq x)$ slowly varying,

$$P(|X_i| \geq x) = o\left(\frac{\lambda(x)}{x^2}\right)$$

$$P(V_n^2 < EV_n^2 - \varepsilon) \rightarrow 0 \quad \star$$

$$(2) P(\sum X_{i,1} \geq (1-\varepsilon)x_n V_n)$$

$$\leq P(\sum X_{i,1} \geq (1-\varepsilon)x_n V_n, V_n \geq (1-\varepsilon)^{\frac{1}{2}} \sqrt{n})$$

$$+ P(\sum X_{i,1} \geq (1-\varepsilon)x_n V_n, V_n < (1-\varepsilon)^{\frac{1}{2}} \sqrt{n})$$

Benette Hoeffding's Inequality
 $X_i \leq a \quad EX_i \leq a$
 $P(\sum X_i \geq x) \leq \exp\left(-\frac{x^2}{2\sum EX_i^2 + 2ax}\right)$

$$\leq P(\sum X_{i,1} \geq (1-\varepsilon)^{\frac{3}{2}} x_n \sqrt{n})$$

$$\leq \exp\left(-\frac{x_n^2 n (1-\varepsilon)^3}{2n + x_n \sqrt{n} z_n}\right)$$

$$+ P(V_n^2 < n - n\varepsilon)$$

$$(2) x_n \sqrt{n} z_n \leq \varepsilon n$$

$$\leq P(\sum X_i^2 1(|X_i| \leq z_n) \leq n - \varepsilon n)$$

$$\leq P(\sum X_i^2 1(|X_i| \leq z_n)) \leq \sum EX_i^2 1(|X_i| \leq z_n) - \frac{\varepsilon n}{2}$$

$$\leq \exp\left(-\frac{\varepsilon^2 n^2}{8n EX_i^4 1(|X_i| \leq z_n)}\right) \leq \exp(-2x_n^2) \#$$

$$E X_i^4 \mathbb{1}(|X_i| \leq z) = o(x^2) \geq \frac{n}{z_n^2 o(1)} \geq \frac{x_n^2}{O(1)} \quad \text{choose } z_n = \varepsilon \sqrt{n}/x_n$$

as $E X_i^2 < \infty$.

■ $X Y \leq \frac{1}{2} \left(\frac{x^2 b^2 + y^2}{b} \right), \quad b > 0, \quad b = \frac{y}{x}$

$$x_n v_n \leq \frac{1}{2} \left(\frac{b^2 v_n^2 + x_n^2}{b} \right), \quad \text{let } b = \frac{x_n}{v_n}$$

when $E X_i^2 = 1$

$$b_n = \frac{x_n}{\sqrt{n}}, \quad b_n \rightarrow 0$$

$$P(S_n \geq x_n v_n)$$

$$\geq P(S_n \geq \frac{1}{2} \left(\frac{b^2 v_n^2 + x_n^2}{b} \right))$$

$$= P(b S_n - \frac{1}{2} b^2 v_n^2 \geq \frac{x_n^2}{2})$$

$$= P\left(\sum (b X_i - \underbrace{\frac{1}{2} (b X_i)^2}_{\xi_i}) \geq \frac{x_n^2}{2}\right)$$

essential that $b \rightarrow 0$
can estimate mgf then

$\{\eta_i\}$ indep.

$$P(\eta_i \leq y) = \frac{E(e^{\lambda \xi_i} \mathbb{1}(\xi_i \leq y))}{E e^{\lambda \xi_i}}$$

$$P\left(\sum \xi_i \geq \frac{x_n^2}{2}\right) = E\left(e^{\lambda \sum \xi_i}\right)^n \cdot e^{-\lambda \sum y_i} \cdot I\left(\sum y_i \geq \frac{x_n^2}{2}\right)$$

want to control $E(\sum y_i)$ wif
too far away from $x_n^2/2$

$$2 \geq \frac{\sum y_i - E y_i}{\sqrt{\sum \text{Var } y_i}} \geq \boxed{\frac{\frac{x_n^2}{2} - E y_i}{\sqrt{n \text{Var}(y_i)}}} \quad (\text{for } \lambda)$$

$$\geq E\left(e^{\lambda \sum \xi_i}\right)^n e^{-\lambda(\sqrt{2\text{Var} y_i} + \sum E y_i)} \cdot P(2 \geq \dots \geq 2)$$

$$EY=0 \quad \text{Var } Y=1$$

$$P(2 \geq Y_1 \geq 2) = 1 - P(|Y| > 2) \geq 1 - \frac{1}{4} = \frac{3}{4}$$

$$\lambda = 1 \pm \varepsilon, \text{ choose } \lambda \text{ such that } (\text{for } \lambda) < -2$$

$$\text{then } \frac{\sum y_i - E y_i}{\sqrt{\sum \text{Var } y_i}} \text{ bounded.}$$