

X_1, X_2, \dots i.i.d. sequence of r.v.s, all on same probability space (Ω, \mathcal{A}, P) with finite common mean μ .

$$E|X_1| < \infty.$$

$$S_n = \frac{X_1 + \dots + X_n}{n}$$

Weak Law of Large numbers (WLLN):

$$\frac{S_n}{n} \xrightarrow{P} \mu$$

Strong law of large numbers (SLLN):

$$\frac{S_n}{n} \xrightarrow{a.s.} \mu.$$

for WLLN,

we need to prove: $P\left(\left|\frac{S_n}{n} - \mu\right| > \varepsilon\right) \rightarrow 0 \quad \forall \varepsilon > 0.$

$$P\left(\left|\frac{S_n}{n} - \mu\right| > \varepsilon\right) = P(|S_n - n\mu| > n\varepsilon)$$

↓ tail probability of the r.v.: $(S_n - n\mu)$

Assume: 2nd moment finite.

$$\text{i.e., } E|X_1|^2 < \infty$$

$$P\left(\left|\frac{S_n}{n} - \mu\right| > \varepsilon\right) = P(|S_n - n\mu| > n\varepsilon) \leq \frac{E|S_n - n\mu|^2}{n^2 \varepsilon^2}$$

(Chebyshev)

$$= \frac{\text{Var}(S_n)}{n^2 \varepsilon^2}$$

$$= \frac{n \cdot \text{Var}(X_1)}{n^2 \varepsilon^2}$$

↓ pairwise independence

$$\rightarrow 0 \text{ as } n \rightarrow \infty$$

Q: How to prove if only $E|X_1| < \infty$ is given?

i.e., only first moment finite is given.

New idea: "Truncation".

↳ Replace X_n 's by Y_n 's, where

Y_n 's are X_n 's "truncated" at an "appropriate threshold."

Here, define $Y_n := X_n \cdot \mathbb{1}_{|X_n| \leq n}$.

Y_1, Y_2, \dots while are still independent

they are no longer identically distributed, as each Y_i has different truncation levels.

Define $T_n := Y_1 + Y_2 + \dots + Y_n$.

We will show: $\frac{T_n - E T_n}{n} \xrightarrow{P} 0$ ——— (*)

& this will then imply, ? why? *

$$\frac{S_n - n\mu}{n} \xrightarrow{P} 0$$

$$\begin{aligned} * \text{ firstly, } \sum P(X_n \neq Y_n) &= \sum P(|X_n| > n) \\ &= \sum P(|X_1| > n) \quad (\because X_i\text{'s} - \text{iid}) \\ &< \infty. \end{aligned}$$

because: $E|X_1| < \infty$

\therefore By Borel-Cantelli lemma:

$$P(X_n \neq Y_n \text{ for infinitely}) = 0$$

\therefore By Borel-Cantelli lemma:

take complement $\left\{ \begin{array}{l} P(X_n \neq Y_n \text{ for infinitely many } n) = 0 \end{array} \right.$

$$\Rightarrow P(\{\omega: X_n(\omega) = Y_n(\omega) \text{ for all } n \geq n_0(\omega)\}) = 1$$

$$\Rightarrow P\left(\left\{\omega: \left(\frac{1}{n} S_n - \frac{1}{n} T_n\right) \rightarrow 0\right\}\right) = 1.$$

$$\therefore \frac{S_n}{n} - \frac{T_n}{n} \xrightarrow{\text{a.s.}} 0 \quad \text{--- (1)}$$

$$\text{Now, } \frac{T_n - E T_n}{n} \xrightarrow{P} 0 \quad \& \quad \frac{S_n}{n} - \frac{T_n}{n} \xrightarrow{P} 0$$

Sum up these two

$$\frac{T_n}{n} - \frac{E T_n}{n} + \frac{S_n}{n} - \frac{T_n}{n} \xrightarrow{P} 0 + 0$$

$$\Rightarrow \frac{S_n}{n} - \frac{E T_n}{n} \xrightarrow{P} 0 \quad \text{--- (2)}$$

$$\therefore E(Y_n) = E(X_1 \cdot 1_{|X_1| \leq n}) \xrightarrow{\text{DCT.}} \mu$$

$$\frac{E(T_n)}{n} = \frac{E(Y_1) + \dots + E(Y_n)}{n} \rightarrow \mu.$$

(real seq. $\rightarrow \mu$
 \Rightarrow their Cesaro mean $\rightarrow \mu$) --- (3)

* & (3) \Rightarrow

$$\frac{S_n}{n} - \mu \xrightarrow{P} 0.$$

\therefore Now, we have to show: $\frac{T_n - ET_n}{n} \xrightarrow{P} 0$.

Fix $\varepsilon > 0$.

$$\begin{aligned} P\left(\left|\frac{T_n - ET_n}{n}\right| > \varepsilon\right) &= P(|T_n - ET_n| > n\varepsilon) \\ &\leq \frac{\text{Var}(T_n)}{n^2\varepsilon^2} \\ &= \frac{1}{n^2\varepsilon^2} \cdot \sum_{k=1}^n V(Y_k) \quad \leftarrow \text{pairwise independent} \\ &\leq \frac{1}{n^2\varepsilon^2} \cdot \sum_{k=1}^n E|X_k|^2 \cdot 1_{|X_k| \leq k} \end{aligned}$$

We're stuck --

(New trick:
fix any non-ve seq. $\{a_n\} \nearrow \infty$, but $\frac{a_n}{n} \rightarrow 0$,
eg. $a_n = \sqrt{n}$.)

$$\begin{aligned} &\leq \frac{1}{n^2\varepsilon^2} \cdot \left(\sum_{k=1}^n E(|X_k|^2 \cdot 1_{|X_k| \leq a_k}) + \right. \\ &\quad \left. \sum_{k=1}^n E(|X_k|^2 \cdot 1_{a_k \leq |X_k| \leq n}) \right) \\ &\quad \swarrow \text{1st term} \quad \searrow \text{2nd term} \\ &\leq \sum_{k=1}^n a_k \cdot E(|X_k| \cdot 1_{|X_k| \leq a_k}) \\ &\leq \sum_{k=1}^n a_k \cdot E(|X_k|) = \end{aligned}$$

$$\therefore \frac{1}{n^2\varepsilon^2} \cdot (\text{1st term}) = \frac{1}{n^2\varepsilon^2} \cdot \sum_{k=1}^n a_n \cdot E(|X_k|)$$

$$\begin{aligned}\therefore \frac{1}{n^2 \epsilon^2} \cdot (1^{\text{st}} \text{ term}) &= \frac{1}{n^2 \epsilon^2} \cdot \sum_{k=1}^n a_n \cdot E(|X_1|) \\ &= \frac{\cancel{n} a_n \cdot E(X_1)}{n^2 \epsilon^2} \rightarrow 0\end{aligned}$$

& 2nd term,

$$\sum_{k=1}^n E(X_1^2 \cdot 1_{a_n < |X_1| \leq n})$$

$$\leq \sum_{k=1}^n n E(|X_1| \cdot 1_{a_n < |X_1| \leq n})$$

$$\leq \sum_{k=1}^n n \cdot E(|X_1| \cdot 1_{|X_1| > a_n})$$

$$\therefore \frac{1}{n^2 \epsilon^2} (2^{\text{nd}} \text{ term}) = \frac{1}{n^2 \epsilon^2} \cdot \cancel{n} \cdot \sum E(|X_1| \cdot 1_{|X_1| > a_n})$$

$$= \frac{1}{\cancel{n} \epsilon^2} \times \cancel{n} \cdot E(|X_1| \cdot 1_{|X_1| > a_n})$$

$$= \frac{E(|X_1| \cdot 1_{|X_1| > a_n})}{\epsilon^2} \xrightarrow{(D.C.T.)^*} 0$$

$\because X_1$ - finite 1st moment
 $\therefore E|X_1| \leq M < \infty$

Hence, we are done. \square

Claim: for WLLN, we only need to show,

$$P\left(\left|\frac{S_n - ES_n}{n}\right| > \epsilon\right) \rightarrow 0$$

for SLLN, we need to show

$$\sum P\left(\left|\frac{S_n - ES_n}{n}\right| > \epsilon\right) < \infty.$$

$$\sum_n P\left(\left|\frac{S_n - ES_n}{n}\right| > \varepsilon\right) < \infty.$$

Borel's Strong Law of Large Numbers:

→ Proved SLLN under stronger assumption that,
(supposedly) $E|X_1|^4 < \infty$ (4th moments finite)

Denote $Y_n := X_n - EX_n = X_n - \mu$ ← essentially centering the X_i 's.

$$T_n = Y_1 + \dots + Y_n.$$

We have to now show: $\frac{T_n}{n} \xrightarrow{a.s.} 0$.

$$\text{i.e., } \sum_n P\left(\left|\frac{T_n}{n}\right| > \varepsilon\right) < \infty.$$

$$P\left(\left|\frac{T_n}{n}\right| > \varepsilon\right) \leq \frac{E(T_n^4)}{n^4 \varepsilon^4} \quad [\text{Chebyshev.}]$$

$$E(T_n^4) = E\left(\left(\sum_{i=1}^n Y_i\right)^4\right)$$

Y_i 's are independent

$$4 \cdot \sum_{i \neq j} E(Y_i^3 \cdot Y_j) \stackrel{!}{=} 4 \cdot \sum E(Y_i^3) E(Y_j) \stackrel{!}{=} 0$$

$$\text{Now, } E\left(\sum_{i=1}^n Y_i\right)^4 = E\left(\sum_{i=1}^n Y_i^4\right) + 4 \cdot E\left(\sum_{i \neq j} Y_i^3 \cdot Y_j\right) +$$

$$4 \cdot \left(\text{Similarly}\right) + 6 \cdot E\left(\sum_{i \neq j} Y_i^2 \cdot Y_j^2\right)$$

$$= n E(Y_1^4) + 6 \cdot n(n-1) \cdot E(Y_1^2 \cdot Y_2^2)$$

$$= n E(Y_1^4) + 6 \cdot n(n-1) \cdot E(Y_1^2 Y_2^2)$$

$$\therefore P\left(\left|\frac{T_n}{n}\right| > \varepsilon\right) \leq \frac{E(T_n^4)}{n^4 \varepsilon^4} = \frac{n E(Y_1^4) + 6 n(n-1) E(Y_1^2 Y_2^2)}{n^4 \varepsilon^4}$$

↓
of the order $\frac{1}{n^2}$

$$\therefore \sum(\cdot) < \infty \quad \checkmark$$

$$\therefore \frac{S_n}{n} \xrightarrow{a.s.} E|X|.$$

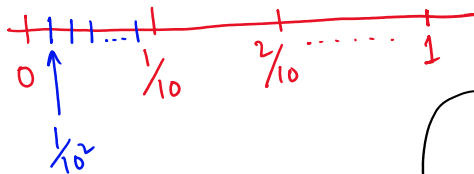
$\Omega = (0, 1]$, $\mathcal{a} = \text{Borel } \sigma\text{-field}$

$P = \text{leb. measure}$

$$X(\omega) = \omega$$

$$X_1(\omega) = \begin{cases} 0, & \text{if } \omega \in (0, \frac{1}{10}] \\ 1, & \text{if } \omega \in (\frac{1}{10}, \frac{2}{10}] \\ \vdots \\ 9, & \text{if } \omega \in (\frac{9}{10}, 1] \end{cases}$$

$$|\omega - X_1(\omega)| \leq \frac{1}{10}$$



$$X_2(\omega) = \begin{cases} 0, & \text{if } \omega - \frac{X_1(\omega)}{10} \in \left(0, \frac{1}{10^2}\right] \\ 1, & \text{if } \omega - \frac{X_1(\omega)}{10} \in \left(\frac{1}{10^2}, \frac{2}{10^2}\right] \\ \vdots \\ 9, & \text{if } \omega - \frac{X_1(\omega)}{10} \in \left(\frac{9}{10^2}, \frac{1}{10}\right] \end{cases}$$

This is basically the non-terminating

This is basically the non-terminating decimal expansion of ω

$$\text{ie, } \omega = 0.X_1(\omega)X_2(\omega)\dots$$

\swarrow decimal
 \downarrow 1st place after decimal
 \downarrow 2nd place after decimal.

Note that, X_i 's are identically distributed.

Q. are they independent?

$$P(X_1=2, X_2=5, X_3=1)$$

$$= P\left(\omega: \omega \in \left(\frac{2}{10} + \frac{5}{10^2} + \frac{1}{10^3}, \frac{2}{10} + \frac{5}{10^2} + \frac{2}{10^3}\right)\right)$$

$$= \frac{1}{10^3} = P(X_1=2) \cdot P(X_2=5) \cdot P(X_3=1)$$

Yes ✓.

\therefore We have i.i.d seq of r.v.s, bounded

\downarrow
 ie, all moments finite.

$$\text{Common mean} = \frac{0+1+2+\dots+9}{10}$$

$$= \frac{9}{2}$$

\therefore By Borel SLLN,

$$\frac{X_1 + \dots + X_n}{n} \xrightarrow{\text{a.s.}} \frac{9}{2}$$

Now, say, $Z_n := 1 \cdot X_n = 3$.

Q. are Z_n 's iid=?
Yes!!

n/2 11

Yes!!

$$\begin{aligned} \therefore P(Z_n=1) \\ = P(X_n=3), \\ \text{ \& } X_n\text{'s are} \\ \text{ i.i.d.} \end{aligned}$$

$$E(Z_n) = 1 \cdot \frac{1}{10} + 0 + 0 + \dots + 0$$

$$\downarrow \\ P(X_n=3)$$

this is the
actually
proportion
of 3's in
the first n
decimal places \rightarrow

$$\left(\frac{Z_1 + Z_2 + \dots + Z_n}{n} \right) \xrightarrow{\text{a.s.}} \frac{1}{10}$$

$$P\left(\omega: \text{For any } i \in \{0, 1, \dots, 10\}, \right. \\ \left. \text{proportion of } \underline{i} \text{ in decimal expansion} \right. \\ \left. \text{of } \omega \rightarrow \frac{1}{10} \right) = 1$$

* Borel's Normal Number Theorem.

\hookrightarrow "almost every no. is normal".

\downarrow
a no. for which,
for every k , the k -adic
expansion of the
that no.,
proportion of every
no. in that
expansion $\rightarrow \frac{1}{k}$