

18 Three Series Theorem

In Probability/Applications one is always interested in understanding how

$$S_n = \sum_{i=1}^n X_i \quad \text{behaves as } n \rightarrow \infty,$$

where $\{X_i\}_{i=1}^\infty$ are i.i.d. r.v. on $(\Omega, \mathcal{F}, \mathbb{P})$.

Example:-

• $X_i \sim \text{Bernoulli}(p)$, then S_n is the number of heads in n -tosses of a coin with probability of heads = p ; $0 \leq p \leq 1$.

• $X_i = \begin{cases} +1 & \text{w.p. } p \\ -1 & \text{w.p. } 1-p \end{cases}$ then S_n is the position of a random walk on $\mathbb{Z}$.

• $X_i \geq 0$ then S_n is called a renewal process
(general distribution)

A famous theorem that deals with sums is due to Kolmogorov. It provides necessary and sufficient conditions for S_n to converge.

They are given in terms of three other series. one assumes that $\{|X_n| > 1\}$ occur finitely often, the second assumes that $E\{X_n | X_n \leq 1\}$ sums up and the third assumes that the variances of $X_n | X_n \leq 1$ sum up

Theorem 1 (Kolmogorov three series theorem) :- If $\{X_n\}_{n \geq 1}$ are independent random variables then

$$S_n = \sum_{k=1}^n X_k \text{ converges almost surely}$$

iff

$$(i) \sum_{n=1}^{\infty} P(|X_n| > 1) < \infty$$

$$(ii) \sum_{n=1}^{\infty} E[X_n \mathbb{1}_{|X_n| \leq 1}] < \infty$$

$$(iii) \sum_{n=1}^{\infty} E[(X_n \mathbb{1}_{|X_n| \leq 1} - E[X_n \mathbb{1}_{|X_n| \leq 1}])^2] < \infty$$

Equivalently

$$Y_i = X_i \mathbb{1}_{|X_i| \leq 1}$$

$$\sum_{n=1}^{\infty} P(X_i \neq Y_i) < \infty$$

$$\sum_{n=1}^{\infty} E[Y_n] < \infty$$

$$\sum_{n=1}^{\infty} \text{var}[Y_n] < \infty$$

Proof:-

- Sufficiency : (i), (ii), (iii) $\Rightarrow \{S_n\}_{n \geq 1}$ converges a.s.

For this we will need the following result:-

Theorem 2 (Khintchine - Kolmogorov) let $\{Y_n\}_{n \geq 1}$ be independent r.v.'s such that

$$E[Y_n] = 0 \quad \forall n \geq 1 \quad \text{and} \quad \sum_{n=1}^{\infty} E[Y_n^2] < \infty$$

then there exist r.v. Y such that

$$\sum_{k=1}^n Y_k \xrightarrow{\text{a.s.}} Y \quad \text{as } n \rightarrow \infty$$

Assuming the result we shall finish the proof of sufficiency (Theorem 1).

$$\text{let } Y_n = X_n \mathbb{1}_{|X_n| \leq 1} - E[X_n \mathbb{1}_{|X_n| \leq 1}] \quad \forall n \geq 1.$$

Then Y_n 's are independent and $E[Y_n] = 0$.

$$\text{Further } \sum_{n=1}^{\infty} E[Y_n^2] < \infty \quad \text{by (iii)}$$

Theorem 2
 $\Rightarrow$

$$\sum_{n=1}^{\infty} Y_n < \infty \quad \text{a.s.}$$

Using (ii) $\Rightarrow \sum_{n=1}^{\infty} X_n \mathbb{1}_{|X_n| \leq 1} < \infty \quad \text{a.s.} \quad \text{--- (1)}$

Now let $Z_n = X_n \mathbb{1}_{|X_n| \leq 1}$

$$\mathbb{P}(X_n \neq Z_n) = \mathbb{P}(|X_n| > 1)$$

By (i) $\sum_{n=1}^{\infty} \mathbb{P}(X_n \neq Z_n) < \infty$

Borel
 $\Rightarrow$
Cantelli

$$\mathbb{P}(X_n \neq Z_n \text{ i.o.}) = 0 \quad \text{--- (2)}$$

$\therefore$ (1)

$$\Rightarrow \mathbb{P}\left(\sum_{n=1}^{\infty} Z_n < \infty\right) = 1$$

By (2)

$$\Rightarrow \mathbb{P}\left(\sum_{n=1}^{\infty} X_n < \infty\right) = 1$$

$$\Rightarrow \mathbb{P}(S_n \text{ converges}) = 1$$

□

• Necessary: S_n converges a.s. to X then (i), (i') & (ii) hold.

We will need the following lemma for proving the necessary direction.

(Two Series Theorem)

Lemma 3:- If Y_n 's are independent uniformly bounded

Y_n 's i.e. $\exists c > 0 \quad \mathbb{P}(|Y_n| \leq c \quad \forall n \geq 1) = 1$ &

• $T_n = \sum_{j=1}^n Y_j$ converges a.s. to T as $n \rightarrow \infty$

then $\sum_{n=1}^{\infty} E[Y_n] < \infty$ and $\sum_{n=1}^{\infty} \text{Var}(X_n) < \infty$.

Assuming the result we shall finish the proof of necessary part. (Theorem 1).

As $S_n = \sum_{k=1}^n X_k$ converges to S a.s. as $n \rightarrow \infty$

$$\Rightarrow X_n \xrightarrow{\text{a.s.}} 0 \text{ as } n \rightarrow \infty$$

$$\Rightarrow \mathbb{P}(\limsup_{n \rightarrow \infty} |X_n| > 1) = 0$$

Borel - Cantelli

$$\Rightarrow \sum_{n=1}^{\infty} \mathbb{P}(|X_n| > 1) < \infty \Rightarrow \text{(i) holds} \quad \text{--- (3)}$$

X_n 's are independent

If $Y_n = X_n \mathbb{1}_{|X_n| \leq 1}$ then we have

$$\mathbb{P}(Y_n \neq X_n) = \mathbb{P}(|X_n| > 1)$$

$$\text{Using (3)} \quad \sum_{n=1}^{\infty} \mathbb{P}(Y_n \neq X_n) < \infty$$

Borel - Cantelli

$$\Rightarrow \mathbb{P}(Y_n \neq X_n \text{ i.o.}) = 0 \quad \text{--- (4)}$$

Using (4) & $S_n = \sum_{k=1}^n X_k$ converges a.s. we have

$$\sum_{k=1}^n Y_k \text{ converges a.s.} \quad \text{--- (5)}$$

As Y_n 's are independent and uniformly bounded by 1 and (5) by lemma 3 we have

$$\sum_{n=1}^{\infty} Y_n^2 < \infty \quad \text{and} \quad \sum_{n=1}^{\infty} \text{Var}(Y_n) < \infty$$

which implies that (ii) and (iii) holds $\square$

We will now prove Theorem 2 and Lemma 3.

Theorem 2 (one series Theorem) (Khintchine - Kolmogorov) let $\{Y_n\}_{n \geq 1}$ be independent r.v.'s such that

$$E[Y_n] = 0 \quad \forall n \geq 1 \quad \text{and} \quad \sum_{n=1}^{\infty} E[Y_n^2] < \infty$$

Then there exist r.v. Y such that

with probability one (i) $T_n = \sum_{k=1}^n Y_k, \quad T_n \xrightarrow{\text{a.s.}} Y \quad \text{as } n \rightarrow \infty,$

(ii) $E[(T_n - Y)^2] \rightarrow 0 \quad \text{as } n \rightarrow \infty$ and $E[Y^2] = \sum_{k=1}^{\infty} E[Y_k^2] < \infty$ } Extra not required for proof of Theorem.

Proof:

Now: $T_n - T_m = \sum_{k=m+1}^n Y_k \quad m \geq n$

let $\varepsilon > 0$ be given

$$P(|T_n - T_m| > \varepsilon) \leq \frac{1}{\varepsilon^2} E[(T_n - T_m)^2]$$

$$= \frac{1}{\varepsilon^2} E\left(\sum_{k=m+1}^n Y_k\right)^2$$

Independence $\Rightarrow$ $E[Y_k] = 0$ $\Rightarrow$
$$= \frac{1}{\varepsilon^2} \sum_{k=m+1}^n E[Y_k^2]$$

By hypothesis that $\sum_{k=1}^{\infty} E[Y_k^2] < \infty$

$$\Rightarrow \sup_{m \geq n} P(|T_n - T_m| > \varepsilon) \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

Proposition 17.1 (i) $\Rightarrow \exists \gamma$ s.t. such that

$$T_n \xrightarrow{p} \gamma \text{ as } n \rightarrow \infty.$$

By Theorem 17.1 (Levy's Theorem)

$$T_m \xrightarrow{a.s.} \gamma \text{ as } m \rightarrow \infty \quad \text{(i) is done}$$

Now by Fatou's lemma

$$0 \leq E|T_k - \gamma|^2 \leq \liminf_{m \rightarrow \infty} E|T_k - T_m|^2$$
$$\stackrel{\text{independence}}{\leq} \sum_{l=k}^{\infty} E[Y_l^2] \quad \forall k \geq 1$$

$$\text{As } \sum_{k=1}^{\infty} E[Y_k^2] < \infty \Rightarrow$$

$$E|T_k - \gamma|^2 \longrightarrow 0 \text{ as } k \rightarrow \infty \quad \text{(ii) is done}$$

- (6)

(iii) let $\varepsilon > 0$ be given.

$$\forall A \in \mathcal{F} \quad E[T_m^2 1_A] \leq 4 [E[Y^2 1_A] + E[(T_m - Y)^2 1_A]]$$

• $\exists m \geq 1$ st $m \geq N$

$$E[(T_m - Y)^2 1_A] \leq E[(T_m - Y)^2] < \varepsilon \quad \text{by (b)}$$

• $\exists \delta > 0$ st $P(A) < \delta \Rightarrow$

$$\max \left\{ E[Y^2 1_A], \max_{m \leq N} E[(T_m - Y)^2 1_A] \right\} < \varepsilon$$

$\Rightarrow \forall m \geq 1 \quad P(A) < \delta \Rightarrow$

$$E[T_m^2 1_A] < 4 [2\varepsilon] = 8\varepsilon$$

$\Rightarrow \{T_n^2\}_{n \geq 1}$ are uniformly integrable.

By Theorem 15.3 : $E[T_m^2] \rightarrow E[Y^2]$ as $m \rightarrow \infty$

$$\Rightarrow E[Y^2] = \lim_{k \rightarrow \infty} \sum_{m=1}^k E[T_m^2] \quad \text{(iii) done} \quad \square$$

(2-Series Theorem)

Lemma 3 :- If Y_n 's are independent uniformly bounded

r.v.'s i.e. $\exists c > 0 \quad P(|Y_n| \leq c \quad \forall n \geq 1) = 1$ &

• $T_n = \sum_{j=1}^n Y_j$ converges a.s. to T as $n \rightarrow \infty$

then - $\sum_{n=1}^{\infty} E[Y_n] < \infty$ and $\sum_{n=1}^{\infty} \text{Var}(Y_n) < \infty$.

Proof :-

let $K > 0$ such that

$$(Ex.) \quad P\left(\sup_{n \geq 1} |T_n| < K\right) > 0$$

$$S: \mathcal{N} \rightarrow \mathbb{R}_+ \cup \{\infty\}$$

$$S = \inf \{ n \geq 1 : |T_n| \geq k \} \quad (P(S=\infty) > 0)$$

$$U_n = T_{\min\{S, n\}} = \sum_{j=1}^n Y_j \mathbf{1}_{\{S \geq j\}}$$

$$\text{As } n \rightarrow \infty, \quad U_n \xrightarrow{a.s.} T_S$$

$$\begin{aligned} \sup_n |U_n| &= \sup_n |T_{\min\{S, n\}} + Y_{\min\{S, n\}}| \\ &\leq k + C \end{aligned}$$

$$\Rightarrow E \sup_n |U_n| < \infty$$

$$\stackrel{\text{D.C.T.}}{\Rightarrow} E[U_n] \longrightarrow E T_S \quad \text{as } n \rightarrow \infty$$

$$\text{Now } E[U_n] = \sum_{j=1}^n P(S \geq j) E[Y_j]$$

$$\{ |T_n| < k : n \leq j \} = \{ S \geq j \} \in \sigma(Y_1, Y_2, \dots, Y_{j-1})$$

$$\Rightarrow E[Y_n] = \frac{E[U_n] - E[U_{n-1}]}{P(S \geq n)}$$

$$b_n = \frac{1}{P(S \geq n)} \quad a_n = E[U_n] - E[U_{n-1}], \quad a_0 = A_0 = 0$$

$$A_n = \sum_{j=0}^n a_j$$

$$\sum_{j=1}^n E[Y_j] = \frac{E[U_n]}{P(S \geq n)} - \sum_{j=1}^{n-1} \left[\frac{1}{P(T \geq j+1)} - \frac{1}{P(T \geq j)} \right] E[U_j]$$

$$\sum_{j=1}^n a_j b_j = A_n b_n - A_0 b_1 - \sum_{j=1}^{n-1} (b_{j+1} - b_j) A_j$$

$$\begin{aligned} \sum a_n < \infty \\ b_n \text{ bounded} \end{aligned} \Rightarrow \sum_{j=1}^{\infty} a_j b_j < \infty$$

$$\Rightarrow \sum_{j=1}^{\infty} E[Y_j] < \infty \quad \Leftrightarrow \quad E[Y_n] \rightarrow E[Y] \quad (P(S=\infty) > 0)$$

Now look at $L_n = \sum_{i=1}^n (Y_i - E[Y_i])$, $L_n \xrightarrow{a.s.} L$

Choose $K_1 > 0$ st. $P(\sup_n |L_n| < K_1) > 0$

$$V = \inf \{ n \geq 1 \mid |L_n| \geq K_1 \}$$

Note: $P(V < \infty) > 0$

$$V_n = \sum_{j=1}^n (Y_j - E[Y_j]) \mathbb{1}_{\{V \geq j\}}$$

$$V_n^2 = V_{n-1}^2 + 2V_{n-1} (Y_n - E[Y_n]) \mathbb{1}_{\{V \geq n\}} + (Y_n - E[Y_n])^2 \mathbb{1}_{\{V \geq n\}}$$

$$E V_n^2 = E V_{n-1}^2 + P(V \geq n) \text{Var}(Y_n)$$

$$\Rightarrow E V_n^2 \leq C_1$$

$$\Rightarrow P(V < \infty) \sum_{j=1}^{\infty} \text{Var}(Y_j) \leq E V_n^2 \leq C_1$$

$$\Rightarrow \sum_{j=1}^{\infty} \text{Var}(Y_j) < \infty \quad \square$$

19 Weak-Convergence

$(\Omega, \mathcal{F}, P)$ be a Probability space. $\{X_n: n \geq 1\}$ and X be random variables on it.

$$\cdot X_n \xrightarrow{P} X \text{ as } n \rightarrow \infty \text{ [Convergence in Probability]}$$

$$\cdot X_n \xrightarrow{a.s.} X \text{ as } n \rightarrow \infty \text{ [Convergence almost surely]}$$

[Proposition 1 and 2 in 17]

Example :-

$$X_n \sim \text{Binomial}(n, \frac{\lambda}{n})$$

$$\cdot \phi_n(k) = P(X_n = k) \xrightarrow{n \rightarrow \infty} \frac{e^{-\lambda} \lambda^k}{k!} \quad \forall k=0,1,2,\dots \quad \text{--- (1)}$$

$$X_n \sim \text{Binomial}(n, \frac{\lambda}{n}) \xrightarrow{n \rightarrow \infty} X \sim \text{Poisson}(\lambda)$$

$$\text{(1)} \Leftrightarrow P(X_n \leq x) \xrightarrow{n \rightarrow \infty} P(X \leq x) \quad x \in \dots$$

$$\cdot X_n \sim \text{Binomial}(n, p)$$

$$Z_n(a, b] = P(a < \frac{X_n - np}{\sqrt{np(1-p)}} \leq b)$$

Binomial C.L.T
(De-Moivre)

$$\xrightarrow{n \rightarrow \infty}$$

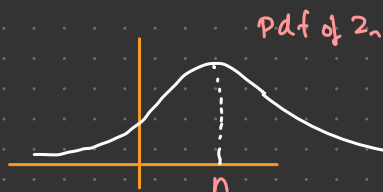
$$\int_a^b \frac{e^{-\frac{z^2}{2}}}{\sqrt{2\pi}} dz \quad \text{--- (2)}$$

|||

$$Z((a, b])$$

(2) $\sqrt{np(1-p)}$.. Scale Binomial (n, p)
 np .. Centre $\xrightarrow{n \rightarrow \infty}$ Normal (μ, σ^2)

• $Z_n \sim \text{Normal}(n, 1)$



$\xrightarrow{n \rightarrow \infty}$ mass "escapes to ∞ ".

Definition 1 :- Given a sequence of probability measures

$\{\mathbb{P}_n(\cdot)\}_{n \geq 1}$ on $(\mathbb{R}, \mathcal{B}_{\mathbb{R}})$ we shall say

" $\mathbb{P}_n(\cdot)$ - Converges weakly to a measure $\mathbb{P}$ "
as $n \rightarrow \infty$

denoted by $\mathbb{P}_n \Rightarrow \mathbb{P}$

if $\int_{\mathbb{R}} f d\mathbb{P}_n \xrightarrow{n \rightarrow \infty} \int_{\mathbb{R}} f d\mathbb{P}$ for all bounded
Continuous $f: \mathbb{R} \rightarrow \mathbb{R}$

Remark 1 :

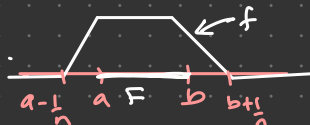
X_n ... law is given by $\mathbb{P}_n$ i.e. $\mathbb{P}(X_n \in \cdot) \equiv \mathbb{P}_n(\cdot)$
 X ... $\mathbb{P}$ i.e. $\mathbb{P}(X \in \cdot) \equiv \mathbb{P}(\cdot)$

" $X_n \xrightarrow[n \rightarrow \infty]{\text{weakly}} X$ " : $\mathbb{E}[f(X_n)] := \int f d\mathbb{P}_n \rightarrow \int f d\mathbb{P} := \mathbb{E}[f(X)]$
 f - bounded continuous f
 $\mathbb{P}_n \Rightarrow \mathbb{P}$ as $n \rightarrow \infty$

- "weak convergence" .. has connections weak topology

- "bounded continuous functions" - captures the topology

of $\mathbb{R}$. i.e. "F-closed set" $1_P(\cdot) \iff f$ bounded continuous



- well-defined notion: $\mathbb{P}_n \Rightarrow \mathbb{P}$ and $\mathbb{P}_n \Rightarrow \mathbb{Q}$
 then $\mathbb{Q} = \mathbb{P}$ (True: force $\int f d\mathbb{P} = \int f d\mathbb{Q}$ for f -bounded continuous)

Theorem 1 (Portmanteau) $\therefore$ let $\{\mathbb{P}_n\}_{n \geq 1}$, $\mathbb{P}$ be Probability

measures on $(\mathbb{R}, \mathbb{B}_{\mathbb{R}})$. The following conditions are equivalent :-

(i) $\mathbb{P}_n \Rightarrow \mathbb{P}$

(ii) $\int f d\mathbb{P}_n \rightarrow \int f d\mathbb{P}$ for all bounded uniformly continuous $f: \mathbb{R} \rightarrow \mathbb{R}$

(iii) $\limsup_{n \rightarrow \infty} \mathbb{P}_n(F) \leq \mathbb{P}(F) \quad \forall F \text{ - closed}$

$$(iv) \quad \lim_{n \rightarrow \infty} P_n(G) = P(G) \quad \forall G \text{ open}$$

$$(v) \quad A \in \mathcal{B}_{\mathbb{R}} \quad P(\partial A) = 0 \quad \text{then} \\ P_n(A) \longrightarrow P(A) \quad \text{as } n \rightarrow \infty.$$

Remark 2:

$$\bullet \quad X_n \sim P_n \quad \text{and} \quad X \sim P$$

$$"X_n \xrightarrow[n \rightarrow \infty]{\text{weakly}} X" \quad (\Leftrightarrow) \quad P_n \Rightarrow P$$

$$\begin{aligned} & \xleftrightarrow{(v)} F_n(x) := P(X_n \leq x) = P(X_n \in (-\infty, x]) \\ & \xrightarrow[n \rightarrow \infty]{} P(X \leq x) := F(x) \end{aligned}$$

$$\text{Ponder: } P(\partial(-\infty, x]) = 0 \quad (\Leftrightarrow) \quad P(\{x\}) = 0 \quad (\Leftrightarrow) \quad x \text{ is a continuity point of } F(\cdot)$$

$$(\text{Alternative: } X_n \xrightarrow{d} X \text{ as } n \rightarrow \infty)$$

$$\bullet \quad P_n(dx) = \text{Uniform} \left\{ \frac{i}{n} : 1 \leq i \leq n \right\}$$

f -bounded continuous

$$\int f dP_n = \sum_{i=1}^n \frac{1}{n} f\left(\frac{i}{n}\right) \xrightarrow[n \rightarrow \infty]{\text{Riemann sum}} \int_0^1 f(x) dx$$

$$\bullet \quad P(dx) = \text{Uniform}(0,1) \quad \int f dP = \int_0^1 f(x) dx$$

$$\boxed{\text{So } P_n \Rightarrow P}$$

(v) - implications

$$A = \bigcap [0, 1]$$

$$\mathbb{P}_n(A) = 1 \quad \& \quad \mathbb{P}(A) = 0$$

(v) - not violated as $\mathbb{P}(\delta A) = \mathbb{P}([0, 1]) = 1$.

- $\{x_n\}_{n \in \mathbb{N}}$ be a sequence in $\mathbb{R}$. Such that $x_n \rightarrow x$ as $n \rightarrow \infty$.

$$\mathbb{P}_n(\cdot) = \delta_{x_n}(\cdot) \quad ; \quad \mathbb{P}(\cdot) = \delta_x(\cdot)$$

Verify Theorem 1 (i) ... (v) :-

(i) f is bounded continuous function.

$$(\mathbb{P}_n \Rightarrow \mathbb{P})$$

$$\int f d\mathbb{P}_n = f(x_n)$$

$$\int f d\mathbb{P} = f(x) \quad \leftarrow \begin{array}{l} n \rightarrow \infty \end{array}$$

(ii) f is bounded uniformly continuous (True from (i))

(iii) F -closed set

$$x \in F \Rightarrow \mathbb{P}(F) = 1 \text{ and clearly}$$

$$\lim_{n \rightarrow \infty} \mathbb{P}_n(F) \leq 1 = \mathbb{P}(F) \quad (\checkmark)$$

$$x \notin F \Rightarrow \mathbb{P}(F) = 0$$

$$F^c - \text{open} \quad \equiv \quad \exists n: n \geq N \quad x_n \notin F.$$

$$x \in F^c$$

$$\exists \varepsilon > 0 \quad B(x, \varepsilon) \subseteq F^c \quad \Rightarrow \quad P_n(F) = 0 \quad \forall n \geq N$$

$$\Rightarrow \quad \lim_{n \rightarrow \infty} P_n(F) \leq 0 = P(F) \quad (\checkmark)$$

(iv) Take Compliments ε ^{use} (iii)

$$(v) \quad A \in \mathcal{B}_{\mathbb{R}} \quad P(\partial A) = 0 \Leftrightarrow x \notin \partial A$$

$$\Leftrightarrow \exists \varepsilon_0 > 0 \quad \text{s.t.} \quad (x - \varepsilon_0, x + \varepsilon_0) \subseteq A \quad \text{or} \quad (x - \varepsilon_0, x + \varepsilon_0) \subseteq A^c$$

$$\Rightarrow \exists N \geq 1 \quad \text{s.t.} \quad n \geq N$$

$$x_n \in (x - \varepsilon_0, x + \varepsilon_0) \begin{cases} \subseteq A \\ \subseteq A^c \end{cases} \quad \text{or}$$

$$\Rightarrow \exists N \geq 1 \quad \text{s.t.} \quad n \geq N$$

$$P_n(A) = 1 = P(A) \quad \text{or}$$

$$P_n(A) = 0 = P(A)$$

$\therefore$ (v) holds

• $\{x_n\}_{n \geq 1}$ differ from x

(iii) $F = \{x \mid \text{inequality is strict}\}$

(v) $A = \{x_n \mid n \geq 1\}$
 $P_n(A)$ has no limit