

MA 203: Convergence of Sequence of RVs

- ✓ 1. ~~Convergence in Mean Square~~
- ✓ 2. ~~Convergence in Distribution~~
3. Convergence in Almost Sure
4. Convergence in Probability

Almost Sure (a. s.) Convergence or Convergence with Probability 1: Let $X_1, X_2, \dots, X_n, \dots$ be a sequence of RVs defined on the probability space (S, F, P) .

The sequence $\{X_n\}_{n=1}^{\infty}$ is said to converge to RV X almost sure or with probability 1 if $P\left(\left\{s \mid \lim_{n \rightarrow \infty} X_n(s) = X(s)\right\}\right) = 1$.

$$s \in S$$

$$A = \left\{s \mid \lim_{n \rightarrow \infty} X_n(s) = X(s)\right\} \quad \varepsilon_n! -$$

$$\begin{matrix} s_1 \\ s_2 \\ \vdots \\ s_m \end{matrix}$$

$$P(A) = 1$$

$$A = \left\{s_i \mid \lim_{n \rightarrow \infty} X_n(s_i) = X(s_i)\right\} \quad S$$

$$t$$

$$s_1, s_2, \dots, s_t$$

$$t \leq m$$

$$\{m-t\}$$

$$\text{if } P(A) = 1 \quad \{X_n\} \xrightarrow{P.} X$$

$$A = \{s_1, s_2, \dots, s_t\}$$

Example 1: Suppose $\underline{S} = \{s_1, s_2, s_3\}$ and $\{X_n\}_{n=1}^{\infty}$ be a sequence of RVs with $X_n(s_1) = 1$, $X_n(s_2) = -1$ and $X_n(s_3) = n$.

Define a RV X such that $X(s_1) = 1$, $X(s_2) = -1$, and $X(s_3) = 1$.

Examine if $\{X_n\}$ converges to X with probability 1 (or almost sure).

Sol:-

$$A = \left\{ s \mid \lim_{n \rightarrow \infty} X_n(s) = X(s) \right\} \quad s \in \underline{S}$$

$$\Rightarrow X_n(s_1) = X(s_1) = 1$$

$$X_n(s_2) = X(s_2) = -1$$

$$X_n(s_3) = n \quad ; \quad X(s_3) = 1$$

$$X_n(s_3) \neq X(s_3)$$

$$A = \{s_1, s_2\}$$

$$\begin{aligned} \text{if } P(A) &= 1 \\ &= P\{s_1, s_2\} \end{aligned}$$

$$\boxed{P(s_3) = 0}$$

Example 2: Let $\{X_n\}$, $n = 1, 2, \dots$, be a sequence of RVs defined on (Ω, F, P) s.t.

$$X_n(s) = 1 + \frac{1}{n}; n = 1, 2, \dots$$

Examine if $\{X_n\}$ converges to $\{X = 1\}$ with probability 1 (or almost sure).

Sol:-

$$A = \left\{ s \mid \lim_{n \rightarrow \infty} X_n(s) = X(1) \right\} \quad \lim_{n \rightarrow \infty} X_n(s) = \lim_{n \rightarrow \infty} 1 + \frac{1}{n} = 1$$

$$= S$$

$$A = S \Rightarrow P(S) = 1$$

Given sequence converges to $\{X=1\}$ with probability 1,
 $\{X_n\} \xrightarrow{\text{a.s.}} \{X=1\}$

$\{X_n\}$

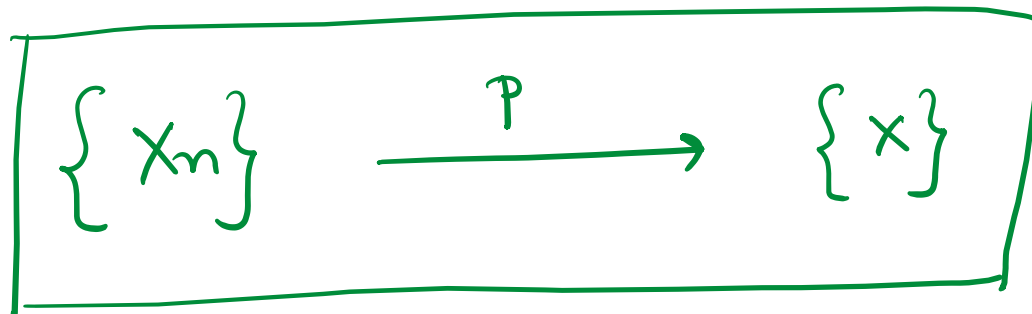
Convergence in Probability

Let $\hat{\{X_n\}}$ be a sequence of RVs defined on (Ω, F, P) . We say that $\{X_n\}, n = 1, 2, \dots$, converges in probability to X If, for any $\epsilon > 0$, $\lim_{n \rightarrow \infty} P\{|X_n - X| > \epsilon\} = 0$.

$$\lim_{n \rightarrow \infty} P\{|X_n - X| > \epsilon\} = 0$$

 $\epsilon > 0$

Notation



Example 3: Let $\{X_n\}$, $n = 1, 2, \dots$, be a sequence of RVs defined on (Ω, F, P) s. t.

$$P(X_n = 0) = 1 - \frac{1}{n} \text{ and } P(X_n = n) = \frac{1}{n}.$$

Examine that $\{X_n\} \rightarrow \{X = 0\}$ in probability.

Sol:-

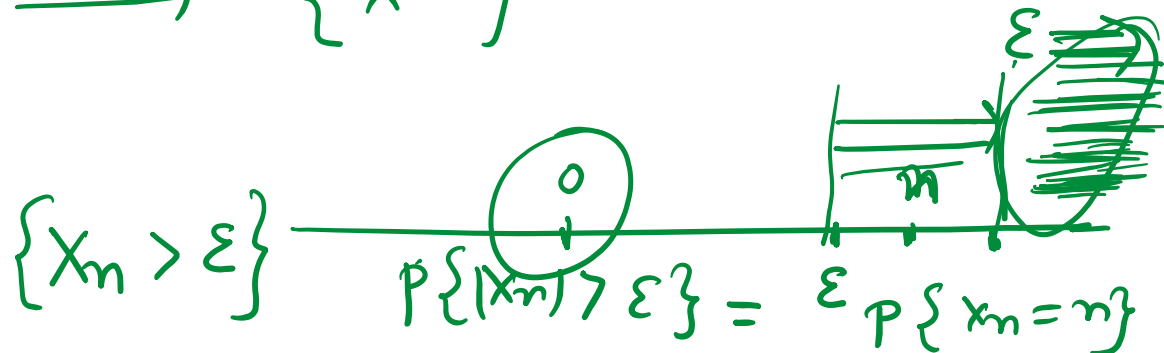
$$\lim_{n \rightarrow \infty} P\{ |X_n - X| > \varepsilon \} = 0$$

$$\Rightarrow P\{ |X_n - 0| > \varepsilon \} =$$

$$P\{ |X_n| > \varepsilon \}$$

$$= \begin{cases} \frac{1}{n} & ; \varepsilon \leq n \\ 0 & ; \varepsilon > n \end{cases}$$

$$\{X_n\} \xrightarrow{P} \{X=0\}$$



$$\lim_{n \rightarrow \infty} P\{ |X| > \varepsilon \} = 0$$

$$\frac{\varepsilon > 0}{\varepsilon > 0}$$

Example 4: Suppose $\{X_n\}$ be a sequence of RVs with

$$P(X_n = 0) = 1 - \frac{1}{n^2}$$

and

$$P(X_n = n) = \frac{1}{n^2}$$

$$\begin{aligned} P\{X_n > 5\} \\ &= P\{\underline{X_n = 10}\} \end{aligned}$$

$$\{X_n\} \xrightarrow{P} X = \dots$$

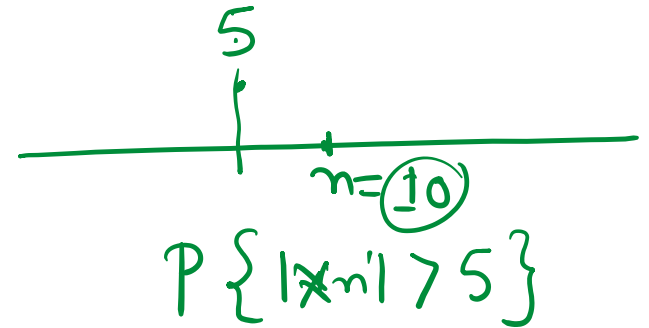
Examine if $\{X_n\}$ converges to $\{X_n = 0\}$ in probability.

Sol: -

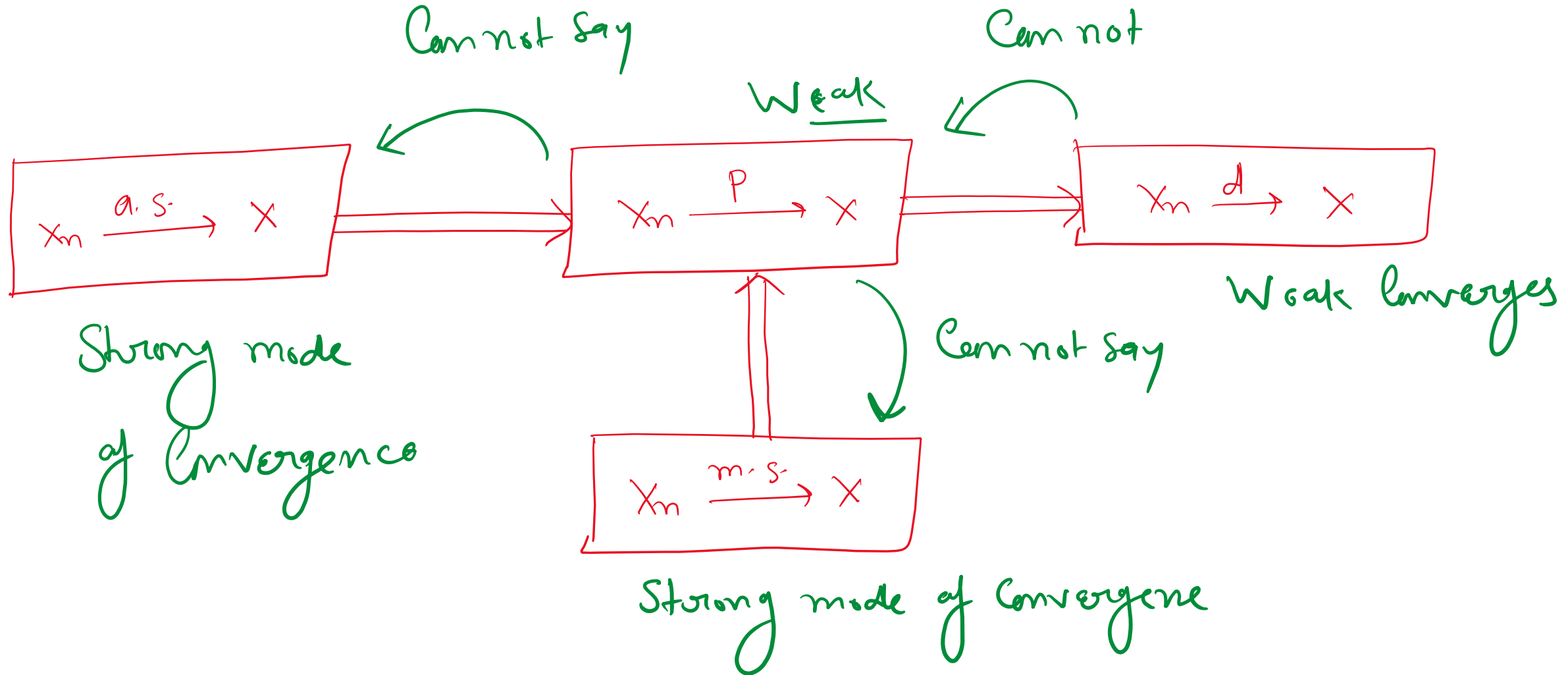
$$\lim_{n \rightarrow \infty} P\{|X_n - X| > \varepsilon\} = 0$$

$$\Rightarrow P\{|X_n - X| > \varepsilon\} = P\{|X_n| > \varepsilon\} = \begin{cases} \frac{1}{n^2} & ; \varepsilon \leq n \\ 0 & ; \varepsilon > n \end{cases}$$

$$\Rightarrow \lim_{n \rightarrow \infty} P\{|X_n| > \varepsilon\} = 0 \quad \{X_n\} \xrightarrow{P} \{X\}$$



Relation Between Different Convergence Modes



Applications:

1. Law of Large Numbers ✓✓
2. Central Limit Theorem ✓✓

Sample Mean: Consider a sequence of RVs $\{X_i\}_{i=1}^{\infty}$ with $\mu_i = E[X_i]$.

The sample mean is defined by

$$\frac{S_n}{n} = \frac{1}{n} \sum_{i=1}^n X_i$$

Where $S_n = \sum_{i=1}^n X_i$.

Then,

$$E\left[\frac{S_n}{n}\right] = \frac{1}{n} \sum_{i=1}^n \mu_i$$

$$E\left[\frac{S_n}{n}\right] = \frac{1}{n} E\left[\sum_{i=1}^n X_i\right]$$

$$= \frac{1}{n} \sum_{i=1}^n \mu_i \Rightarrow$$

True Mean

$$S_n = \sum_{i=1}^n X_i$$

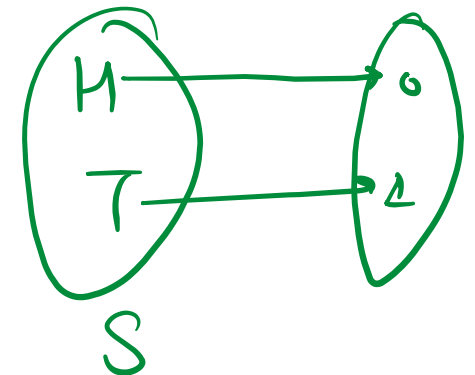
$$= X_1 + X_2 + \dots + X_n$$

$X_1 \quad X_2 \quad \dots \quad X_n \quad \dots \quad X_M$

$X_1 + X_2 + \dots + X_n$

$$\frac{0 + 1 + \dots + 1}{n} =$$

Sample mean



μ_n $n > M$