

# Convergence of Random Variables

1. Convergence in Probability: A sequence  $X_1, X_2, \dots$  of r.v. converges in prob. to another r.v.  $Y$  if:  $\forall \varepsilon > 0, \lim_{n \rightarrow \infty} P(|X_n - Y| \geq \varepsilon) = 0$ .  
 $\Leftrightarrow \forall \varepsilon > 0, \lim_{n \rightarrow \infty} P(|X_n - Y| < \varepsilon) = 1$ .

written as  $\{X_n\} \xrightarrow{P} Y$

e.g.  $Y \sim \text{Uniform}[0, 1]$  and  $X_n = (1 + \frac{1}{n})Y$ . Does  $X_n \xrightarrow{P} Y$ ?

Let  $\varepsilon > 0$ . WTS:  $\lim_{n \rightarrow \infty} P(|X_n - Y| < \varepsilon) = 1$ .

$$|X_n - Y| = |(1 + \frac{1}{n})Y - Y| = \frac{1}{n}Y \leq \frac{5}{n} < \varepsilon \Rightarrow n > \frac{5}{\varepsilon}, \text{ which.}$$

$\forall n > \frac{5}{\varepsilon}, P(|X_n - Y| < \varepsilon) = 1$ , which.  $\lim_{n \rightarrow \infty} P(|X_n - Y| < \varepsilon) = 1$ . i.e.  $X_n \xrightarrow{P} Y$ .

1. In general,  $\forall \varepsilon > 0, P(|X_n - Y| \geq \varepsilon) \leq P(X_n \neq Y)$ . which. if  $\lim_{n \rightarrow \infty} P(X_n \neq Y) = 0$ . then  $\{X_n\} \xrightarrow{P} Y$ .

e.g.  $U \sim \text{Uniform}[0, 1]$ .  $X_n = \mathcal{I}_{U \leq \frac{1}{2} + \frac{1}{2n}}$ .  $Y = \mathcal{I}_{U \leq \frac{1}{2}}$ . Does  $X_n \xrightarrow{P} Y$ ?

Let  $\varepsilon > 0$ .  $P(|X_n - Y| \geq \varepsilon) \leq P(X_n \neq Y)$

Since.  $\frac{U_1, U_2, U_3}{\frac{1}{2}, \frac{1}{2} + \frac{1}{2n}, \frac{1}{2} + \frac{1}{2n}} \rightarrow$  only  $U = U_2$  satisfies  $X_n \neq Y$ .

$$\text{which is } P(\frac{1}{2} < U \leq \frac{1}{2} + \frac{1}{2n}) = \frac{\frac{1}{2n}}{1-0} = \frac{1}{2n}.$$

Since  $\lim_{n \rightarrow \infty} P(X_n \neq Y) = \lim_{n \rightarrow \infty} \frac{1}{2n} = 0$ , gives.  $X_n \xrightarrow{P} Y$

2. Weak Law of Large Numbers: For any sequence of r.v.  $X_1, X_2, \dots$  which are independent, each have the same  $\mu$ , and each have variance  $\leq v$  for  $v < \infty$ , if  $M_n = \frac{1}{n}(X_1 + X_2 + \dots + X_n)$ , then  $M_n \rightarrow \mu$  in probability. i.e.  $\lim_{n \rightarrow \infty} P(|M_n - \mu| \geq \varepsilon) = 0, \forall \varepsilon > 0$ .

proof: From Chebychev's inequality.  $P(|M_n - \mu| \geq \varepsilon) \leq \frac{\text{Var}(M_n)}{\varepsilon^2}$ .

when  $n \rightarrow \infty, \frac{v}{n\varepsilon^2} \rightarrow 0$  as needed. which.  $\leq \frac{v}{n\varepsilon^2}$

$M_n \xrightarrow{P} \mu$

1)  $\{X_n\}$  are i.i.d., i.e., independent and identically distributed.

$\rightarrow$  All have same  $\mu$  &  $\sigma^2$ .  $\rightarrow$  All have the same Prob.

when its not identity func. try use the method as math class.

when its identical func. use the generalized property.

$$\text{Var}(M_n) \leq \frac{v}{n}$$

$$E(M_n) = \mu$$

If  $\{X_n\}$  i.i.d., then no need cond. ③.

$P(a \leq X_n \leq b)$  is same for all  $n$ .

e.g.  $X_n =$   $n^{\text{th}}$  coin Heads.

(Bernoulli).

$\{X_n\}$  indep. with  $E(X_n) = \frac{1}{2} = \mu$ .  $\text{Var}(X_n) = \frac{1}{4} \cdot (1 - \frac{1}{2}) = \frac{1}{8} < \infty$ .

if  $M_n = \frac{1}{n}(X_1 + X_2 + \dots + X_n)$  is fraction of Head on first  $n$ . by WLLN,  $M_n \xrightarrow{P} \mu$ , i.e.  $\lim_{n \rightarrow \infty} P(|M_n - \mu| \geq \varepsilon) = 0, \forall \varepsilon > 0$ .

Can't directly apply to  $X_n$  as for  $0 < \varepsilon < \frac{1}{2}$ ,  $P(|X_n - \frac{1}{2}| \geq \varepsilon) = 1$ .  
 $\downarrow$   
 $X_n = 0 \text{ or } 1$ .

Part EG.

3. Convergence Almost Surely: A sequence  $X_1, X_2, \dots$  of r.v. C.A.S. to another r.v.  $Y$ , if  $P(X_n \rightarrow Y \text{ as a sequence}) = 1$ , i.e.  $P(\lim_{n \rightarrow \infty} X_n = Y) = 1$ , written as  $X_n \xrightarrow{\text{a.s.}} Y$ .  
 $\leftarrow$  def in math.  $\forall \varepsilon > 0, \exists n \geq N$  s.t.  $|X_n - Y| < \varepsilon$ .

1). If  $X_n \xrightarrow{\text{a.s.}} Y$ , then  $X_n \xrightarrow{P} Y$ .

e.g.  $P(X_n = 7) = \frac{1}{9 \cdot 10^{k-1}}$ ,  $P(X_n = 5) = 1 - [\frac{1}{9 \cdot 10^{k-1}}]$ , i.e. one 7 in  $X_1$  to  $X_9$ , one 7 in  $X_{10}$  to  $X_{99}$ , one 7 in  $X_{100}$  to  $X_{999}, \dots$

①.  $X_n \xrightarrow{P} 5$ .

$P(|X_n - 5| \geq \varepsilon) \leq P(X_n \neq 5) = P(X_n = 7)$ , when  $n \rightarrow \infty, P(X_n = 7) \rightarrow 0$ .

$\therefore \{X_n\} \xrightarrow{P} 5$ .

②.  $X_n \xrightarrow{\text{a.s.}} 5$ .

$\downarrow$  def in math.  
 $X_n \rightarrow 5$  as a sequence, i.e.  $\forall \varepsilon > 0, \exists n \geq N$  s.t.  $|X_n - 5| \leq \varepsilon$ .

However, we always have  $X_n = 7$  which  $|X_n - 5| = 2 > \varepsilon$ .

① Intuition:  $\forall \varepsilon > 0$ , as  $n \rightarrow \infty$ .

$\rightarrow X \xrightarrow{P} Y: P(|X_n - Y| \geq \varepsilon) \rightarrow 0$ .

$\rightarrow X \xrightarrow{\text{a.s.}} Y: P(\exists m \geq n \text{ with } |X_m - Y| \geq \varepsilon) \rightarrow 0$ .

4. Strong law of large numbers (SLLN): For any sequence of r.v.

$X_1, X_2, \dots$  which are ① i.i.d., each with same ②  $\mu$ , if  $M_n = \frac{1}{n}(X_1 + \dots + X_n)$ .

then  $M_n \xrightarrow{\text{a.s.}} \mu$

e.g.  $X_n = \text{1st coin Heads}$ .

$\{X_n\}$  i.i.d. with  $\mu = E(X) = \frac{1}{2}$ .

① By WLLN.  $\forall \varepsilon > 0$ .  $\lim_{n \rightarrow \infty} P(|M_n - \frac{1}{2}| \geq \varepsilon) = 0$ . i.e.  $P(M_n < 0.503) > 0.99$ .

② By SLLN.  $P(M_n \rightarrow \frac{1}{2}) = 1$ . i.e.  $P(\exists n: M_n < 0.503) = 1$ .

$\Rightarrow P(\exists n: X_1 + \dots + X_n < 0.503n) = 1$ .

