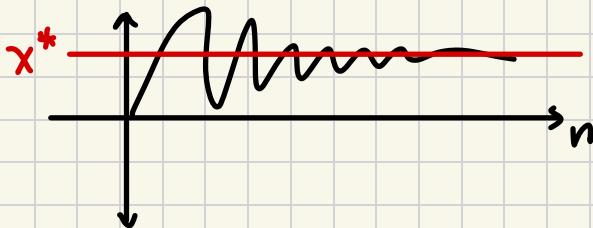


# Lec 3: Convergence of Random Variables

## Convergence of sequences

$X_1, X_2, X_3, \dots$

$$\lim_{n \rightarrow \infty} X_n = X^* \quad X_n \xrightarrow{n \rightarrow \infty} X^*$$



$X_1, X_2, X_3, \dots$

$X_n \rightarrow X$

① *almost sure*

$X_n(w) \rightarrow X(w) \quad \forall w \in \Omega$

*(almost all)  
w ∈ Ω*

② the pdf of  $X_n$

$\downarrow$  conv.

the pdf of  $X$ .

$\Omega$  sample space

$\mathbb{P}$

$X : \Omega \rightarrow \mathbb{R}$

$w \in \Omega$

$\Omega, X_i : \Omega \rightarrow \mathbb{R}$

(imagine each  $X_i$  is a feature of something (like  $X_1 = \text{temp}$ ,  $X_2 = \text{date}$ , ...))

$w \in \Omega$

↑ specifies the values

"all at once"

now RV values map to  $\mathbb{R}$

ex:  $X_1, X_2, \dots$  are iid,  $\mathbb{E}[X_i] = \mu$ ,  $\mathbb{V}[X_i] = \sigma^2$

$$\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i \quad \bar{X}_n \rightarrow \mu \quad (\text{SLLN})$$

↳ strong law...

### Convergence in probability:

def:  $X_n \xrightarrow{\text{IP}} X$  if  $\forall \varepsilon > 0$

$$\mathbb{P}(|X_n - X| > \varepsilon) \xrightarrow{n \rightarrow \infty} 0$$

↳ takes whole sample space into account.

if  $n = \text{large}$ , look at  $X_n$  &  $X$  & see how far apart they are.

ex:  $\bar{X}_1, \bar{X}_2, \dots$ ,

$X_n \xrightarrow{\text{iid}} \text{Ber}(\gamma_2)$  → coin toss

$X \sim \text{Ber}(\gamma_2)$

$X$  and  $\{\bar{X}_n\}$  are independent

$\bar{X}_n \xrightarrow{\text{IP}} X ?$

for  $\varepsilon > 0$ , (how to choose  $\varepsilon$  wisely bc  $X \in [0, 1]$ )

$$\mathbb{P}(|\bar{X}_n - X| > \varepsilon) = \mathbb{P}(\{\bar{X}_n = 0\} \cap \{X = 1\}) \cup \{\bar{X}_n = 1\} \cap \{X = 0\}$$

↳ proper notation

$$= \mathbb{P}(X_n = 0, X = 1) + \mathbb{P}(X_n = 1, X = 0) \leftarrow \text{bc disjoint}$$

$$= \mathbb{P}(X_n = 0) \mathbb{P}(X = 1) + \mathbb{P}(X_n = 1) \mathbb{P}(X = 0)$$

$$= \gamma_2 \cdot \gamma_2 + 1 - \gamma_2 \cdot \gamma_2 = \gamma_2 \neq 0$$

(WLLN) exercise prove this for  $X_n$  iid w/  $\mathbb{E}[X_n] = \mu$

weak

Markov

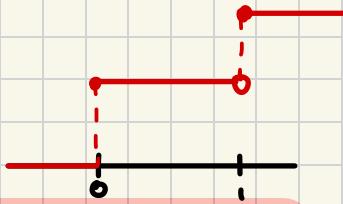
$$\mathbb{V}[X_n] = \sigma^2 < \infty$$

## Convergence in dist:

$X_n \rightsquigarrow X$

$$F_X(x) = \mathbb{P}(X \leq x)$$

↑ CDF of  $X$



If CDF of  $X_n$  converges to the CDF of  $X$  at all continuity points of CDF of  $X$ .

ex: (CLT) (central limit thm) can show dist. of var.  
 $\text{Bin}(n, 1/2)$  (take gaussian of CDF & compare)

how to write:  $\Rightarrow$  or  $\xrightarrow{(d)}$  or  $\xrightarrow{d}$  or  $\xrightarrow{\mathcal{L}}$  or  $\mathcal{L}(x_n) \rightarrow \mathcal{L}(x)$

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Thm: if  $X_n \xrightarrow{P} X$  then  $X_n \rightsquigarrow X$ .

Lem: if  $X_n \rightsquigarrow c$  (constant) ↓ conv. in dist.  
then  $X_n \xrightarrow{P} c$

## exercise

let  $\varepsilon > 0$ :

$$\mathbb{P}(|X_n - c| > \varepsilon) > \dots$$

\*  $\frac{e^{\tan \frac{1}{n^2}}}{\cos \frac{1}{n}} \rightarrow 1$

## Operation:

- if  $X_n \xrightarrow{P} X, Y_n \xrightarrow{P} Y:$

$$X_n + Y_n \xrightarrow{P} X + Y$$
$$X_n Y_n \xrightarrow{P} X Y$$

-  $X_n \rightsquigarrow X$  then  $X_n + Y_n \rightsquigarrow X + c$

$$Y_n \xrightarrow{P} c \quad X_n Y_n \rightsquigarrow cX$$

-  $Y_n = -X_n$

$$X, X_1, X_2, \dots \stackrel{iid}{\rightsquigarrow} N(0,1)$$

$$X_n \rightsquigarrow X$$

$$Y_n \rightsquigarrow X$$

can be anything  
with same dist. as  $X$ , so  $X_1, -X$ , etc.

## Continuous Mapping Thm:

$$X_n \xrightarrow{P} X$$
$$g(X_n) \xrightarrow{P} g(X)$$
$$\rightsquigarrow (\bar{X}_n)^2 ?$$

reading  
↑

\* go thru  
lec notes  
(didn't  
finish  
in lec)

## Delta Method:

$$\frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}} \rightsquigarrow \frac{N(0,1)}{z} \rightarrow \left( \frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}} \right)^2 \rightsquigarrow z^2$$

(not good bc want)

$$\frac{(\bar{X}_n)^2 - 0}{0} \rightsquigarrow 0$$