

- Asymptotic Analysis (a set of "approximation" results)

The objectives ① When do we have "good" properties of an estimator?

→ General conditions are better.

- ② Compare Estimators and decide which ones are better than the others.

(ex) Under basic assumptions \oplus Normality,

OLS is the best unbiased estimator.

Likewise, the asymptotic property makes estimators better.

* Approximation concepts need to be defined.

Starting point: Sequence of numbers converges to a number

$$\lim_{n \rightarrow \infty} x_n = x_{\infty} \text{ if and only if } \forall \varepsilon > 0, \exists N_0: \forall n \geq N_0: |x_n - x_{\infty}| < \varepsilon$$



We want to extend the convergence concept to a sequence of random variables or more generally, a sequence of random vectors.

We want to define a concept of random variables, so that a sequence of random variables can be understood easily.

Example) Throwing a dice infinite times

$$\{1, 2, \dots, 6\} \times \{1, 2, \dots, 6\} \times \{1, 2, \dots, 6\} \times \dots$$

→ Any probability to a number realization is equal to 0.

→ Think of a case: a set is realized as a number.



$$\{1, 2, \dots, 6\} \times \{1, 2, \dots, 6\} \times \{1, 2, \dots, 6\} \times \dots$$

↓
 X_1

↓
 X_2

↓
 X_3

$$X_1=1 \text{ w/p } \frac{1}{6}$$

$$X_1=2 \text{ w/p } \frac{1}{6}$$

⋮

$$X_1=6 \text{ w/p } \frac{1}{6}$$

$$X_2=1 \text{ w/p } \frac{1}{6}$$

⋮

⋮

Considering such a structure
as a sequence,

↓

$$\{X_1, X_2, X_3, \dots\}$$

Then, it is defined as a sequence written by $X_1(\omega), X_2(\omega), \dots$

That is,

For each ω , $\{X_1(\omega), X_2(\omega), X_3(\omega), \dots\}$

⇒ If we fix ω , then the sequence is realized.

If we fix the other $\omega' \neq \omega$, then the sequence is changed.

↳

$X_i : \Omega \rightarrow \mathbb{R}^d$, random variable is a function.

↑
probability

↑
sample space.

In the above case (dice), the distribution is uniform.

Dice	1	2	3	5	1	6	5	5	4	...	<Sequence>
	↓	↓	↓	↓	↓	↓	↓	↓	↓		
corresponding to	0	1	2	4	0	5	4	4	3	...	

$$\text{prob } \left(\frac{1}{6}\right) \cdot 0 + \left(\frac{1}{6}\right)^2 \cdot 1 + \left(\frac{1}{6}\right)^3 \cdot 2 + \left(\frac{1}{6}\right)^4 \cdot 4 + \dots$$

→ This will correspond to
a number in $[0, 1]$

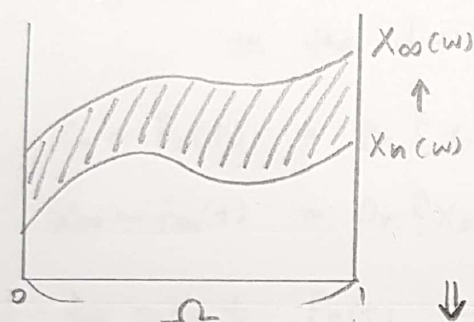
→ $\omega \in \Omega$ can be understood as an infinite data.

$X_1(\omega)$ tells us what we compute using ω , upto 1st observation...
 then $X_n(\omega)$ " " " , upto the n th observation

→ Convergence concept of random variables requires convergence concept of functions.

* \exists 4 modes of convergence.

1. Convergence in r th Mean



$$X_i: \Omega \rightarrow \mathbb{R}^d \text{ (d=1 case)}$$

$$\text{When } E \{ |X_n(\omega) - X_{\infty}(\omega)|^2 \}^{\frac{1}{2}} \rightarrow 0,$$

We can say "convergence in 2nd mean"

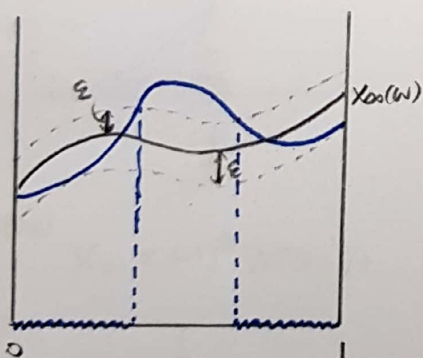
↓ Extend r th moment!

$$\lim_{n \rightarrow \infty} E \|X_n - X_{\infty}\|^r \rightarrow 0$$

i.e., $\{X_n\}$ converges in r th mean to X_{∞}

$$\text{i.e., } X_n \xrightarrow{r} X_{\infty} \Leftrightarrow X_n - X_{\infty} \xrightarrow{r} 0$$

2. Convergence in Probability



Choose an arbitrary ϵ .

If $X_n(\omega)$ is within $X_{\infty}(\omega)$, the case that the sum of probability is going to 1 is convergence in Prob.

That is,

$$\forall \epsilon > 0, \Pr \{ |X_n(\omega) - X_{\infty}(\omega)| < \epsilon \} \rightarrow 1 \text{ if } n \rightarrow \infty$$

$$\text{i.e., } X_n \xrightarrow{P} X_{\infty} \Leftrightarrow X_n - X_{\infty} \xrightarrow{P} 0$$

[3] Almost sure convergence

$\{X_n\}$ converges almost surely to X_∞

$$\text{i.e., } \Pr \left\{ \lim_{n \rightarrow \infty} X_n(\omega) = X_\infty(\omega) \right\} = 1 \quad \text{where } \omega \in \Omega$$

$$\text{i.e., } X_n \xrightarrow{\text{a.s.}} X_\infty \iff X_n - X_\infty \xrightarrow{\text{a.s.}} 0$$

[4] Convergence in distribution

Convergence of the distribution function of X_n to the distribution function of X_∞

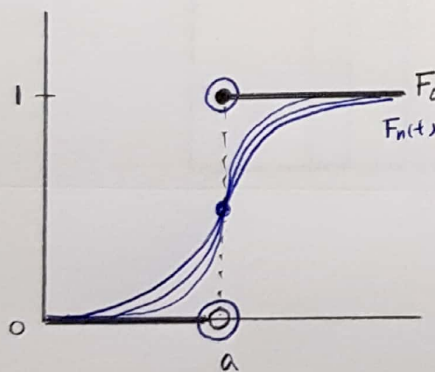
$$X_n \sim F_n(t) = \Pr \{X_n \leq t\}$$

$$X_\infty \sim F_\infty(t) = \Pr \{X_\infty \leq t\}$$

If, for each t , $F_n(t) \rightarrow F_\infty(t)$ as $n \rightarrow \infty$

at the continuity point of $F_\infty(t)$,

then X_n converges in distribution to X_∞



In this case, $F_\infty(t)$ discontinues at a .
However, "At all continuous point of $F_\infty(t)$ "

$$F_n(t) \rightarrow F_\infty(t)$$

This means $F_n(t)$ converges in distribution of $F_\infty(t)$.

ex)

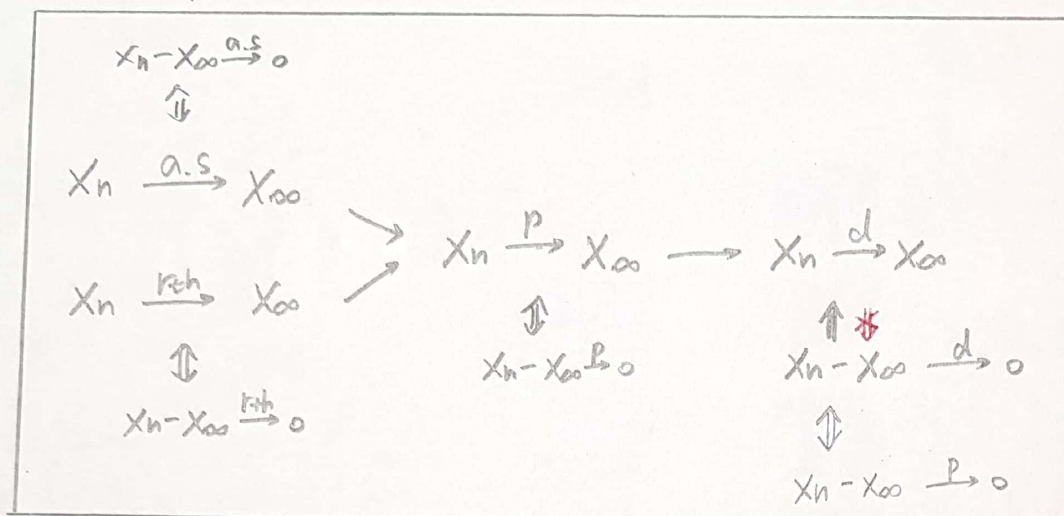
$$X_n = (-1)^n \cdot N(0, 1)$$

↓

$$X_\infty \sim N(0, 1)$$

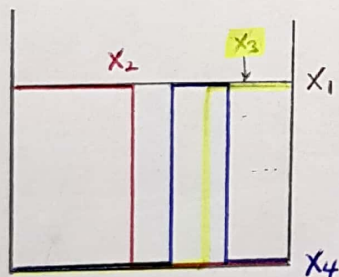
⌋ Their realized value are not the same.
But their distribution can be equal.

⇒ Relationships of 4 modes of convergence.



(Note) X_{∞} is a constant number.
 $X_n \xrightarrow{p} X_{\infty} \Leftrightarrow X_n \xrightarrow{d} X_{\infty}$

Example) $X_n \xrightarrow{a.s.} X_{\infty} \not\Leftarrow X_n \xrightarrow{p} X_{\infty}$



When $n \rightarrow \infty$,

$X_n \xrightarrow{a.s.} X_{\infty}$: all X_n will go to X_{∞}

$X_n \xrightarrow{p} X_{\infty}$: many of X_n will go to X_{∞} ,
 but not guaranteed for all //