

SDS 387 Linear Models

Fall 2024

Lecture 5 - Tue, Sep 10, 2024

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■ Convergence in distribution (weakest form of stochastic convergence)

A sequence of random vectors in \mathbb{R}^d $\{X_n\}$ converges in distribution (or weakly) to a random vector X when, $\forall x \in \mathbb{R}^d$ s.t. F_x is continuous at x ,
cdf of X

$$\left. \begin{array}{l} X_n \xrightarrow{d} X \\ X_n \Rightarrow X \\ X_n \rightsquigarrow X \end{array} \right\} \begin{array}{l} \text{Same} \\ \leftarrow \\ \text{cdf of } X_n \end{array} \quad F_{X_n}(x) \rightarrow F_X(x) \text{ as } n \rightarrow \infty.$$

[pointwise convergence of $\{F_{X_n}\}$ at all continuity points of F_X]

Remarks

1) This definition does not impose any restriction on the joint distribution of the X_n 's and of X

For example $X_n = (-i)^n Z$ $Z \sim N(0,1)$ (1)

and $X \sim N(0,1)$ then $X_n \xrightarrow[\text{w.p.1}]{d} X$
 but $X_n \not\stackrel{p}{\rightarrow} X$ nor $X_n \not\rightarrow X$

i.) The fact that X_n and X have similar cdf's does not mean they take on similar values. In above example $X_n \sim N(0,1)$ and $X \sim N(0,1)$ but in general $|X_n - X|$ can be large

Examples: Let Φ be the cdf of $N(0,1)$ and let

$$i) \quad F_n(x) = \begin{cases} 0 & x < -n \\ \frac{\Phi(x) - \Phi(-n)}{\Phi(n) - \Phi(-n)} & -n \leq x \leq n \\ 1 & x \geq n \end{cases}$$

Then $F_n \rightarrow \Phi$ pointwise

$$ii) \quad F_n(x) = \begin{cases} 0 & x < -n \\ \Phi(x) & -n \leq x < n \\ 1 & x \geq n \end{cases}$$

neither continuous nor discrete

$F_n \rightarrow \Phi$

The restriction that convergence takes place at all continuity points of F_X only is necessary!

Example

$$X_n = \begin{cases} 1 - \frac{1}{n} & \text{wp } 1/2 \\ 0 - \frac{1}{n} & \text{wp } 1/2 \end{cases} \quad n \text{ odd}$$

$$\begin{cases} 1 + \frac{1}{n} & \text{wp } 1/2 \\ 0 + \frac{1}{n} & \text{wp } 1/2 \end{cases} \quad n \text{ even}$$

It is clear that, for large n , $X_n \approx \text{Bernoulli}(1/2)$
 So we would like to say that $X_n \xrightarrow{d} \text{Bernoulli}(1/2)$

That is in fact the case. But for $x=1$

$$F_{X_n}(1) = \begin{cases} 1 & n \text{ odd} \\ 1/2 & n \text{ even} \end{cases}$$

$\hookrightarrow F_{X_n}(1)$ does not converge!!

That is ok: $x=1$ is a point of discontinuity of the cdf $\text{Bernoulli}(1/2)$

Result: If $X_n \xrightarrow{p} X$ then $X_n \xrightarrow{d} X$.

PA / $X_n \xrightarrow{p} X$ means that, $\forall \varepsilon > 0$, $\xrightarrow{\text{small}}$

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

So, for any $x \in \mathbb{R}$,

$$\mathbb{P}(X \leq x - \varepsilon) = \mathbb{P}(\underbrace{\{X \leq x - \varepsilon\} \cap \{|X_n - X| \leq \varepsilon\}}_{\subseteq \{X_n \leq x\}}) +$$

$$\mathbb{P}(\underbrace{\{X \leq x - \varepsilon\} \cap \{|X_n - X| > \varepsilon\}}_{\subseteq \{X_n - X > \varepsilon\}})$$

$$P(A) = P(A \cap B) + P(A \cap B^c)$$

$$\leq \mathbb{P}(X_n \leq x) + \mathbb{P}(|X_n - X| > \varepsilon)$$

$$(*) \quad \hookrightarrow \mathbb{P}(X \leq x - \varepsilon) - \mathbb{P}(|X_n - X| > \varepsilon) \leq \mathbb{P}(X_n \leq x)$$

Similarly,

$$(**) \quad \mathbb{P}(X_n \leq x) \leq \mathbb{P}(X \leq x + \varepsilon) + \mathbb{P}(|X_n - X| > \varepsilon)$$

Because $\liminf \mathbb{P}(|X_n - X| > \varepsilon) = \limsup \mathbb{P}(|X_n - X| > \varepsilon) = 0$
 Inequalities (*) and (**) imply

$$\mathbb{P}(X \leq x - \varepsilon) \leq \liminf_n \mathbb{P}(X_n \leq x) \leq \limsup_n \mathbb{P}(X_n \leq x) \leq \mathbb{P}(X \leq x + \varepsilon)$$

as $\varepsilon \downarrow 0$ goes to

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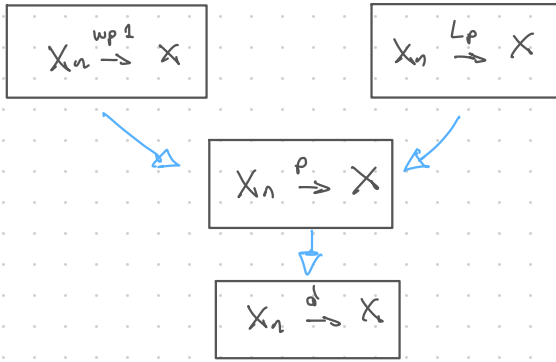
$$\underbrace{\mathbb{P}(X < x)}_{F_X(x^-)} \leq \liminf_n \mathbb{P}(X_n \leq x) \leq \limsup_n \mathbb{P}(X_n \leq x) \leq \underbrace{\mathbb{P}(X \leq x)}_{F_X(x)}$$

If x is a continuity point of F_X , $F_X(x^-) = F_X(x)$

so,

$$\lim_{n \rightarrow \infty} F_n(x) = F_X(x)$$

Schematic of stochastic convergence:



Result:

If $X_n \xrightarrow{d} X$ where X is degenerate (non-random)
 then $X_n \xrightarrow{P} X$

Prf/ Let call $X = a$ [i.e. $\mathbb{P}(X=a)=1$] Then
 we want to show that $\mathbb{P}(|X_n - a| \geq \varepsilon) \rightarrow 0$
 $\forall \varepsilon > 0$

For $\varepsilon > 0$ ^{→ small} Then

$$\mathbb{P}(|X_n - a| \geq \varepsilon) \leq \mathbb{P}(X_n \leq a - \varepsilon) + \underbrace{\mathbb{P}(X_n \geq a + \varepsilon)}_{1 - \mathbb{P}(X_n < a + \varepsilon)}$$

$$\leq \mathbb{P}(X_n \leq a - \varepsilon) + 1 - \underbrace{\mathbb{P}(X_n \leq a + \frac{\varepsilon}{2})}_{\rightarrow 1}$$

$$\rightarrow 0$$

→ 0

Remarks:

joint weak convergence implies weak convergence of marginals but not the other way around

If $\begin{bmatrix} X_n \\ Y_n \end{bmatrix} \xrightarrow{d} \begin{bmatrix} X \\ Y \end{bmatrix}$ then we can conclude

that $X_n \xrightarrow{d} X$ and $Y_n \xrightarrow{d} Y$ HWR!

however, if $X_n \xrightarrow{d} X$ and $Y_n \xrightarrow{d} Y$, in general we cannot conclude that

$$\begin{bmatrix} X_n \\ Y_n \end{bmatrix} \xrightarrow{d} \begin{bmatrix} X \\ Y \end{bmatrix}$$

Example

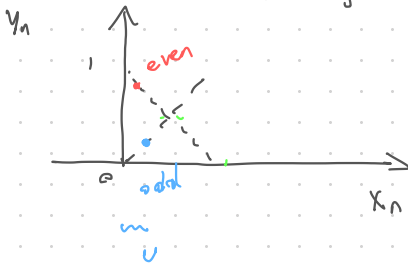
Let $U \sim \text{Uniform}(0,1)$

$$X_n = U \quad \text{all } n$$

$$Y_n = \begin{cases} U & n \text{ odd} \\ 1-U & n \text{ even} \end{cases}$$

Clearly $X_n \xrightarrow{d} U$ and $Y_n \xrightarrow{d} U$

But $\begin{bmatrix} X_n \\ Y_n \end{bmatrix}$ does not converge at all!



HWR

1.2) Let F_n be the cdf of $\text{Uniform}[-n, n]$

Does F_n converge? No!

The issue is that the mass is too

(6)

About v): when $\overset{\text{dimension}}{\downarrow} d=1$ take $A = (-\infty, x]$

then $\mathbb{P}(X_n \in A) = F_n(x)$ and

$$\mathbb{P}(X \in \partial A) = \mathbb{P}(X = x)$$

which is $= 0$ when x is a continuity point of F_x

More generally: X_n is a discrete uniform on $\{0, 1, \dots, n-1\}$

Then $\frac{X_n}{n} \xrightarrow{d} U$ where $U \sim \text{Uniform}(0,1)$

$A = [0, 1] \cap \mathbb{Q}_n$ then $\mathbb{P}(X_n \in A) = 1$

but $\mathbb{P}(U \in A) = 0$

Φ : explain why condition v) is not violated!

HU

Continuous mapping theorem:

Let $f: \mathbb{R}^d \rightarrow \mathbb{R}$ and let X be a random vector in \mathbb{R}^d s.t. $\mathbb{P}(X \in C) = 1$ where C is the set of continuity points of f . Then

$$X_n \xrightarrow{*} X \implies f(X_n) \xrightarrow{*} f(X)$$

where $*$ means up 1 or in probability or in distribution.

Note: if f is continuous this is trivially true