

Last time

$X_n \xrightarrow{D} X$  if  $\underbrace{F_{X_n}(a)}_{\substack{\uparrow \\ \text{CDF}}} \rightarrow F_X(a)$  whenever  $F_X$  is cts at  $a$ .

Levy's continuity theorem: If  $\underbrace{\varphi_{X_n}(t)}_{\substack{\downarrow \\ \text{characteristic ftn}}} \rightarrow \varphi(t)$ , and  $\varphi$  is cts at  $t=0$ , then  $\varphi(t) = \varphi_X(t)$  for some  $X$  and  $X_n \xrightarrow{D} X$

Law of Large # (One of Many)

Let  $X_n$  be a sequence of i.i.d., and  $E(X_n) = b$ .

$$\text{let } Y_n = \frac{1}{n} \sum_{k=1}^n X_k$$

Theorem:  $Y_n \xrightarrow{D} b$ , " $b$ " constant ftn

Proof- Look at  $\varphi_{Y_n}(t) = \varphi(t) \rightarrow$  the characteristic ftn of  $X_n$

$$\text{Then: } \varphi_{Y_n}(t) = E e^{itY_n} = E e^{it \left( \frac{X_1 + \dots + X_n}{n} \right)}$$

$$\begin{aligned} & \stackrel{\text{indep}}{\downarrow} = E \left( e^{\frac{it}{n} X_1} \times \dots \times e^{\frac{it}{n} X_n} \right) \\ & = E \left( e^{\frac{it}{n} X_1} \right) \times \dots \times E \left( e^{\frac{it}{n} X_n} \right) = \varphi \left( \frac{t}{n} \right)^n \end{aligned}$$

Taylor Approximation:

$$\varphi(t) = \varphi(0) + \varphi'(0)t + \delta(t)t, \text{ where } \delta(t) \rightarrow 0 \text{ as } t \rightarrow 0$$

$$\varphi(0) = 1, \varphi'(0) = iE(X_n) = ib$$

$$\text{So } \varphi_{Y_n}(t) = \left( 1 + \frac{ibt}{n} + \underbrace{\delta\left(\frac{t}{n}\right) \frac{t}{n}}_{\text{very small}} \right)^n$$

$$\left( 1 + \frac{ibt}{n} \right)^n \xrightarrow{n \rightarrow \infty} e^{ibt} \quad \text{since } \lim_{n \rightarrow \infty} \left( 1 + \frac{x}{n} \right)^n = e^x$$

$\uparrow$   
characteristic ftn of random variable equal to  $b$  everywhere.

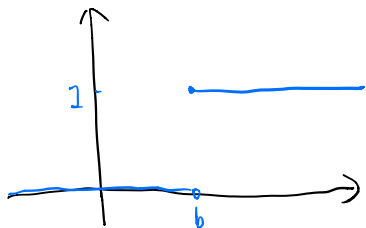
Levy's Theorem:  $Y_n \xrightarrow{D} b$

$$\left( 1 + \frac{ibt}{n} \right)^n$$

$$\lim_{n \rightarrow \infty} \left( 1 + \frac{x}{n} \right)^n = e^x$$

$$\left( 1 + \frac{ibt}{n} \right)^n \rightarrow e^{ibt}$$

CDF of  $b$ :  $F(a) = P(b \leq a) = \begin{cases} 1 & \text{if } a \geq b \\ 0 & \text{if } a < b \end{cases}$



$F_n(a) \rightarrow F(a)$  for all  $a \neq b$ .  
 $\uparrow$   
 CDF of  $Y_n$

$\forall \epsilon > 0, \lim_{n \rightarrow \infty} P(Y_n \geq b + \epsilon) = 0,$

$\lim_{n \rightarrow \infty} P(Y_n \leq b - \epsilon) = 0$

So  $P(Y_n = b)$  all bets are off.

$X_i = \begin{cases} 1 & P = \frac{1}{2} \\ -1 & P = \frac{1}{2} \end{cases}$

$Y_n = \frac{1}{n} \sum_{k=1}^n X_k$

$E Y_n = b = 0$

$P(Y_n = 0) = \begin{cases} 0 & \text{if } n \text{ is odd} \\ 2^{-n} \binom{2n}{n} & \text{if } n=2m \text{ is even} \end{cases}$   
 $\approx \frac{1}{\sqrt{n}}$

How fast the convergence is?

Chebyshev, Bernstein, and others are about that

They require more than just having expectations.

Why do we need  $\psi(t)$  cts at  $t=0$  in Levy's continuity theorem?

Here is what can happen:

Suppose that  $X_n$  is Gaussian with  $E(X_n) = 0$  and  $\text{Var}(X_n) = n$ .

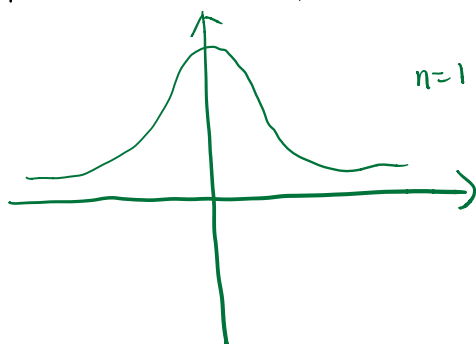
$\psi_{X_n}(t) = e^{-nt^2/2} \xrightarrow{n \rightarrow \infty} \begin{cases} 0 & \text{if } t \neq 0 \\ 1 & \text{if } t = 0 \end{cases}$

So  $\psi(t) = \begin{cases} 0 & \text{for } t \neq 0 \\ 1 & \text{for } t = 0 \end{cases}$

• not cts at  $t=0$

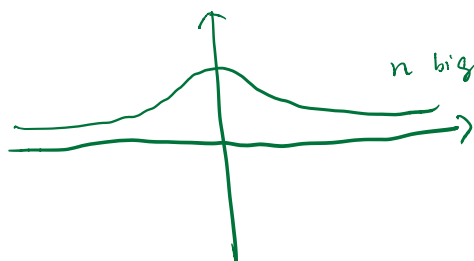


$X_n \xrightarrow{D}$  doesn't converge anywhere in distn.



$n=1$

graph for pdf

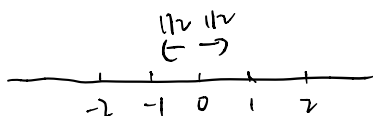


$n$  big (variance is big)

The condition that  $\phi(t)$  is cts at  $t=0$  prevents the distn from "disappearing" in the limit.

## Towards Brownian Motion

Intuition: Consider symmetric random walk starting at  $x=0$  without barriers.



$$X = X_1 + \dots + X_n, \quad X_k = \begin{cases} 1 & p=1/2 \\ -1 & p=1/2 \end{cases}$$

↑  
position at  
the  $n$ -th step

Now: Make each step  $\frac{1}{\sqrt{n}}$  (instead of 1)

And make each step in  $\frac{1}{n}$  second (instead of 1 second)

Position after  $n$  steps:  $Y_n = \frac{1}{\sqrt{n}} \sum_{k=1}^n X_k \xrightarrow{D} N(0,1)$

# Central Limit Theorem

Now try to describe the whole movement.

## Wiener Process

$X_t: \Omega \rightarrow \mathbb{R}$  r.v.  $t \geq 0$

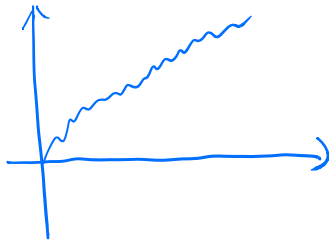
①  $X_0 = 0$

② If  $0 \leq t_1 < t_2 < \dots < t_n$

then random variables:  $X_{t_2} - X_{t_1}, X_{t_3} - X_{t_2}, \dots, X_{t_n} - X_{t_{n-1}}$  are indep. } indep increments  
intuition:  $X_{t_k} - X_{t_{k-1}}$

③ For each  $0 \leq t_1 < t_2$ , the r.v.  $X_{t_2} - X_{t_1}$  is Gaussian with displacement from time  $t_{k-1}$  to  $t_k$   
 $E(X_{t_2} - X_{t_1}) = 0$  and  $\text{Var}(X_{t_2} - X_{t_1}) = t_2 - t_1$

④  $P\{\omega \in \Omega : \text{the fcn } t \mapsto X_t(\omega) \text{ is cts}\} = 1$



## Theorem (Wiener, Levy)

Wiener process exists.

Theorem:  $P\{\omega \in \Omega : \text{the fcn } t \mapsto X_t(\omega) \text{ is not differentiable at any } t\} = 1$

Sanity check:  $X_{t_3} - X_{t_1} = (X_{t_3} - X_{t_2}) + (X_{t_2} - X_{t_1})$   
indep

$$\text{Var}(X_{t_3} - X_{t_1}) = \text{Var}(X_{t_3} - X_{t_2}) + \text{Var}(X_{t_2} - X_{t_1})$$

$$t_3 - t_1 = (t_3 - t_2) + (t_2 - t_1)$$