

## Lecture 5 : Invariance Principle for RCM

Theorem (Y<sub>k</sub>) RW on RCM,  $P^W[Y_{k+1} = y | Y_k = x] = \frac{w_{x,y}}{\pi(x)}$ .

(w) iid with elliptic condition, then

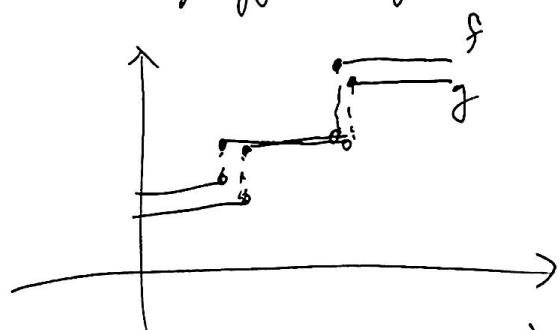
$$P\text{-a.s. } Y_t^{(n)} = \begin{cases} \frac{1}{\sqrt{n}} Y_{[nt]} & \text{not } t \in N \\ \text{interpolation} & \text{otherwise} \end{cases}$$

$$(Y_t^{(n)})_{t \geq 0} \longrightarrow (\beta_t)_{t \geq 0} \text{ in } C([0, \infty))$$

Remark Last time, we talked about Markov Jump Processes, which VSRW, CSRW, which are continuous-time.

This theorem is true for those as well, with Skorokhod topology.

This topology is for ~~continuous~~ or càdlàg functions.



If we use ~~a~~ usual norm for these function, then  $d(f, g) \rightarrow 0$ .

D<sub>f</sub> (Corrector Method).

We don't have a classical CLT, which is why this is difficult. How can we show that the scaling limit is BM?

We want to construct a martingale, which should converge more closely to BM (characterized by continuous martingale with quadratic variation  $t$ ) - Lévy's theorem).

Step 1 (Corrector):  $(\phi_j)_{j=1, \dots, d}$  s.t.  $e_j \cdot x + \underbrace{\phi_j(x)}_{Y_j(x)}$  is harmonic under  $\mathbb{P}_f$ . That is,  $\Psi_j(x)$

$$\left( L f(x) - \sum_{y \neq x} \frac{w_{x,y} (f(y) - f(x))}{\pi(x)} \right)$$

$$L \Psi_j(x) = 0.$$

$\Rightarrow \Psi_j(Y_k)$  is a martingale.

$$\begin{aligned} & \mathbb{E}[\Psi_j(Y_{k+1}) - \Psi_j(Y_{k-1}) \mid \mathcal{F}_{k-1}] \\ &= L \Psi_j(Y_{k-1}) = 0. \end{aligned}$$

Step 2: Martingale convergence.

$$\frac{1}{\sqrt{n}} \Psi_j(Y_{n+1}) \xrightarrow{\text{Convergence in entries}} (\alpha B_t)_{t \geq 0} \quad (\text{Chacon-Wald})$$

Step 3:  $\sup_{[0, T]} \frac{1}{\sqrt{n}} |\phi_j(Y_{n+1})| \xrightarrow{\mathbb{P}} 0$   $\Rightarrow$  Convergence in vector.

Step 3: sufficient condition: multilinearity of corrector.

$$\text{i.e., } \frac{1}{\sqrt{n}} \max_{B \in \mathbb{R}^{d \times d}} |\phi_j(x)| \xrightarrow[n \rightarrow \infty]{\text{Hausdorff}} 0. \quad (1)$$

$$\text{(or } \frac{1}{N} \max_{B \in \mathbb{R}^{d \times d}} |\phi_j(x)| \xrightarrow[N \rightarrow \infty]{} 0)$$

We are trying to correct a linear function to harmonic, so of course the corrector should not be the dominant term. Otherwise, we are not 'correcting', we are changing/replacing.

If  $\phi_j(x)$  linear, then  $\max_{B_{\sqrt{n}L}} |\phi_j(x)| \sim \sqrt{n} \notin o(\sqrt{n})$ .

~~the last~~ So we want  $\sqrt{n}$  to dominate.

Def (informal). This  $Y_{[nt]}$   $\sim BM$ . So in  $(0, \infty)$ ,

$Y_{[nt]}$  should be around  $\sqrt{[nt]}$ . So

~~$\sup_{[0,T]}$~~  the value of  $Y_{[nt]}$  should be similar to the value in the ball  $B_{\sqrt{n}L}$  for some  $L$ . So it should be quite natural for us to try to prove ~~so~~ this. (1)

Def We will use a stopping time argument.

Let  $T_{n,L} = \inf \{ k : Y_k \notin \text{Ball} \}$ .

$$P \left[ \sup_{[0,T]} \frac{1}{\sqrt{n}} |\phi_j(Y_{[nt]})| > \delta \right]$$

$$= P \left[ \dots \text{ } T_{n,L} \leq T \right] + P \left[ \dots \text{ } T_{n,L} > T \right].$$

The second term is easier. It means when  $Y_{[nt]}$  is still in  $B_{\sqrt{n}L}$  the maximum must be less than that on the ball.

$$P \left[ \dots \mid T_{n,L} > T \right] \leq P \left[ \max_{B_{\sqrt{n}L}} \left| \frac{1}{\sqrt{n}} \phi_j(x) \right| > \delta \right] \xrightarrow{n \rightarrow \infty} 0 \quad \text{Actually,}$$

the this convergence is  $P-a.s.$ .

The approximation of the first term requires a theorem about martingales.

$$\bar{\psi}_j(Y_k) = e_j \cdot Y_k + \phi_j(Y_k)$$

$$\frac{1}{\sqrt{n}} \bar{\psi}_j(Y_{nt}) = \frac{1}{\sqrt{n}} e_j \cdot Y_{nt} + \frac{1}{\sqrt{n}} \phi_j(Y_{nt}). \quad (2)$$

If  $T_{n,L} \leq T$ , the our RW is quite ~~not~~ fast.

It reaches the boundary of Box quite quickly.

But as we know, the corrector is small at the boundary. Because when  $n$  is large, at the moment when the RW reaches the boundary, the contribution from the corrector is small so the martingale part is quite large. So if we estimate the martingale, then we can control the corrector.

$$P\left(\sup_{[0,T]} \left| \frac{1}{\sqrt{n}} \phi_j(Y_{nt}) \right| < \delta, T_{nf} \leq T\right)$$

$$\leq P\left[\dots T_{n,L} \leq T, \frac{1}{\sqrt{n}} \max_{B\sqrt{n}L} |\phi_j| \geq 1\right]$$

$$+ P\left[\dots \dots \dots \dots \leq 1\right]$$

The former term also  $\rightarrow 0$ . Why? Because

still look at the large

Notice (\*). we let  $[nt] = T_{n,L}$ . Then what happens?  $\frac{1}{\sqrt{n}} \phi_j(Y_{T_{n,L}}) \leq 1$  for large  $n$ .

We also know  $\sup_{t \in [0, T]} |Y_{[nt]}| \approx 1$ .

So we know the size of the martingale.

$$\mathbb{P} \left[ \sup_{[0, T]} \left| \sum_{j=1}^L \bar{\psi}_j(Y_{[nt]}) \right| \geq L - 1 \right]$$

$$\leq \frac{1}{L-1} \mathbb{P}[\dots]$$

$$\xrightarrow{L \rightarrow +\infty} 0.$$

Dobrinskij inequality

Basically we truncate to add the information about if  $Y_{[nt]} \in \text{BmL}$ .

So we proved a property of the corrector. But we have not talked about its existence ...

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Now, we talk about some martingale convergence theorem thing.

Step 2: If we have a martingale, we create a prob. space, BM, and stopping times. Then BM/ $\sigma$ (stopping times) is the same as our martingale.

Then (Embedding).

Let  $(M_n)$  be a  $L^2$ -martingale. Then  $\exists$  BM  $(B_t)_{t \geq 0}$ ,

and stopping times  $T_k$ .

$$(M_n)_{n \in \mathbb{N}} \stackrel{d}{=} (B_{T_k})_{k \in \mathbb{N}}.$$

Why is this an embedding? We embed a martingale into a BM.

Idea We do a scaling transformation.

$$\left( M_{nt}^{(n)} \right) = \left( \frac{1}{\sqrt{n}} M_{[nt]} \right)_{t \geq 0} \stackrel{d}{=} \left( \frac{1}{\sqrt{n}} B_{T_{[nt]}} \right)_{t \geq 0}.$$

Then in some sense, we want to show that this converges in law to  $(B_t)_{t \geq 0}$ . Is  $T_{[nt]}$  close to  $nt$ ? This is ~~standard~~ scaling invariance of BM.

Very again. We have a martingale. We know that it can be embedded into the path of a BM. Now, where. This lecture is very.

(This theorem from Durrett.)

e.g. Define  $X$ ,  $\mathbb{E}[X] = 0$ ,  $X \in \{-a, 1\}$ .  $a, b > 0$

$$\Rightarrow \mathbb{P}[X = b] = \frac{a}{a+b}, \mathbb{P}[X = -a] = \frac{b}{a+b}.$$

We first construct  $\mathbb{E}[X]$  from BM.  $\mathbb{E}[(-a \wedge b)]$

$$= \mathbb{E}[B_{(-a \wedge b)}^2]$$

$$= \mathbb{E}[X^2] = ab.$$

$$B_{(-a \wedge b)} \stackrel{d}{=} X.$$

e.g. General  $X$ ,  $\mathbb{E}[X] = 0$ ,  $\text{Var}[X] < +\infty$ .

$$C = \int_0^{+\infty} x dF_X(x) = - \int_{-\infty}^0 x dF_X(x).$$

Given  $\ell$  a test function,  $C \mathbb{E}[\ell(X)] =$

$$= \int_{-\infty}^0 \int_0^{+\infty} \ell(x) dF_X(x) dF_Y(y) + \int_0^{+\infty} \int_{-\infty}^0 \ell(y) dF_X(x) dF_Y(y)$$

$$= \int_{-\infty}^0 -y \int_0^{+\infty} \varphi(x) dF_X(x) dF_X(y) + \int_0^{+\infty} x \int_{-\infty}^0 \varphi(y) dF_X(y) dF_X(x)$$

$y \in (-\infty, 0]$   
 $\Rightarrow x \in (0, +\infty)$

$$= \int_{-\infty}^0 \int_0^{+\infty} \left( \frac{x-y}{x+y} \varphi(y) - \frac{y}{x+y} \varphi(x) \right) dF_X(x) dF_X(y)$$

$$\Rightarrow \mathbb{E}[\varphi(X)] = \int_{-\infty}^0 \int_0^{+\infty} \frac{(x-y)}{c} \underbrace{\left[ \frac{x}{x+y} \varphi(y) + \frac{y}{x+y} \varphi(x) \right]}_{\text{2-point distribution}} dF_X(x) dF_X(y).$$

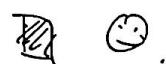
$\frac{x-y}{c} \sim \text{density function}$

$\mathbb{E}[\varphi(B_{T_k} \wedge T_x)]$

Of course, talking about stopping time requires us to include  $X$  sampling  $X$  and  $y$  into filtration. This is not in the natural filtration.

Def for  $(M_n)$ . First we construct  $M_{1,2}$ .

Now, conditioning on  $M_1$ , construct  $M_2$ , etc.



Def (Actual, maybe).

Given  $X$  r.v. with  $\mathbb{E}[X] = 0$ ,  $\mathbb{E}[X^2] < +\infty$ .

We know how to construct:  $B_T \stackrel{d}{=} X$ .

Suppose we have  $(B_{T_1}, \dots, B_{T_{k-1}}) \stackrel{d}{=} (M_1, \dots, M_{k-1})$ .

We let  $\mathbb{P}[X \in \cdot] = \mathbb{P}[M_k \in \cdot | M_1, \dots, M_{k-1}]$ .

Define  $X$  r.v.  $\xrightarrow{\text{to centre}}$  sample an r.v.

$B_{T_{k-1}+T_k} - B_T \stackrel{d}{=} X \stackrel{d}{=} M_k - M_{k-1}$  conditioned on  $M_1, \dots, M_{k-1}$ .

Strong Markov Property:  $\underbrace{\delta_{T_{k-1}+t}}_{:= T_k} - \beta_t$  is BM.

Then (Martingale CLT).

(M<sub>k</sub>)  $L^2$ -martingale,  $\Delta M_k = M_k - M_{k-1}$

If ①  $V_n = \sum_{k=1}^n \mathbb{E}[(\Delta M_k)^2 | \mathcal{F}_{k-1}]$

(quadratic variation)

$$\lim_{n \rightarrow \infty} \frac{V_n}{n} = \sigma^2$$

- ②  $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^n \mathbb{E}[(\Delta M_k)^2 | \mathcal{F}_k] \geq \lim_{n \rightarrow \infty} \left[ \frac{V_n}{n} \right] \xrightarrow{n \rightarrow \infty} 0$ .

$$g_t^{(n)} := \begin{cases} \frac{1}{n} M_{nt} & n \in \mathbb{N} \\ \text{interpolation} & \text{otherwise} \end{cases}$$

↗  
increment cannot be too large.

$$(s_t^{(n)})_{t \geq 1} \xrightarrow{n \rightarrow \infty} (\beta_t)_{t \geq 0} \text{ in } C([t_0, +\infty))$$

We will use the Embedding Theorem.

$$\Delta T_k = T_k - T_{k-1} = \mathbb{E}[(\beta_{T_k} - \beta_{T_{k-1}})^2 | \mathcal{F}_{T_{k-1}}]$$

$$\begin{aligned} \frac{1}{n} M_{nt} &\xrightarrow{n \rightarrow \infty} \mathbb{E}[T_k - T_{k-1} | \mathcal{F}_{T_{k-1}}] \\ \frac{1}{n} \beta_{T_{nt}} &\xrightarrow{n \rightarrow \infty} \mathbb{E}[(M_k - M_{k-1})^2 | \mathcal{F}_{k-1}] \\ \beta_t &\xrightarrow{a.s.} \mathbb{E}[(\Delta M_k)^2 | \mathcal{F}_{k-1}] \end{aligned}$$

The inverse of stopping times  $\Rightarrow$  approx. linear  
I gave up.

## Step 1: Construction of corrector.

There will be a lot of details which will be left to be gradually filled in in later lectures.

Today, we will talk about the construction and ~~and~~ sublinearity will be left for later.

There are a few methods / ways for the construction, each with their own pros and cons.

Def  $(\phi_j)_{j=1,\dots,d}$  s.t.  $L_v(e_j \cdot x + \phi_j(x)) = 0 \forall x \in \mathbb{R}^d$ .  
we don't really care about its uniqueness (if any) since  
we just want a martingale

The difficulty is the  $(\forall x \in \mathbb{R}^d)$  part.

\* If we only require this for  $x \in \text{Br } \mathbb{R}^d$ . This is easy. We can just resolve the discrete harmonic equation in  $\text{Br}$ .  $\rightsquigarrow \phi_j^{(n)}(x)$ .

\* Why is unbounded domain problematic? We want to look at this equation in a function space. If we put this in bounded spaces, then the linear function is ~~is~~ not in the space  $H(\mathbb{R}^d)$ .

Is it possible that  $\phi_j^{(n)} \xrightarrow{n \rightarrow +\infty} \phi_j$  in some sense?  
This is a good idea which will appear throughout later parts of this course. This is quite a PDE approach.

Another method is just to find the function. But it's weird sort of sick what he is saying.

## Construction by random field

"derivative about environment"

$$\nabla_x \varphi(\omega) := \varphi(T_x \omega) - \varphi(\omega), \quad \varphi: \Omega \rightarrow \mathbb{R}^d$$

We ~~drag~~ the random environment a bit and take the difference.

potential field.  $L_{p.t.}^2$ , with finite  $\langle \cdot, \cdot \rangle$

$$\langle \varphi, \varphi \rangle = \mathbb{E} \left[ \sum_x \omega_{0,x} (\nabla_x \varphi(\omega))^2 \right]$$

Similar to Dirichlet energy.

Then  $\exists !$  minimizer for the problem,

$$\varphi \mapsto \mathbb{E} \left[ \sum_x \omega_{0,x} |x + \nabla_x \varphi(\omega)|^2 \right] \quad (x)$$

over shift ~~-invariant~~ covariant field.

$$F: \Omega \times \mathbb{Z}^d \rightarrow \mathbb{R}^d.$$

$$\text{Shift covariance: } F(\omega, x+z) - F(\omega, x)$$

$$= F(T_x \omega, z)$$

$$x+z-z=x$$

$$\nabla_{x+z} \varphi(\omega) - \nabla_x \varphi(\omega)$$

$$= \varphi(T_{x+z} \omega) - \varphi(T_x \omega)$$

$$= \nabla_z \varphi(T_x \omega).$$

Ok, we do this optimisation problem. Then the minimiser with

Def The functional is positive. So if  $\varphi = 0$ , then already has some upper bound on the value. So existence ✓.

"Uniqueness": & Make it unique with equivalence relation.

Let  $\phi \neq \emptyset$  be minimiser of  $(*)$ . Then for any  $\varphi: S \rightarrow \mathbb{R}^d$ , locally  $L^2$ ,  $\frac{d}{d\varphi} \mathbb{E} \left[ \sum_x w_{0,x} \|x + \nabla \varphi(\omega) + \varepsilon \nabla_h \varphi(\omega)\|^2 \right] = 0$ .

$$\Rightarrow \cancel{\langle \phi, \varphi \rangle} = 0.$$

$$\mathbb{E} \left[ \sum_x w_{0,x} \underbrace{\langle \nabla \varphi(\omega), (x + \nabla \varphi(\omega)) \rangle}_{\varphi(T_x \omega) - \varphi(\omega)} \right] = 0.$$

$$\mathbb{E} \left[ \sum_x w_{0,x} \underbrace{\varphi(T_x \omega) \langle x + \nabla \varphi(\omega) \rangle}_{\|x\|} \right] - \mathbb{E} \left[ \underbrace{\varphi(\omega) \langle x + \nabla \varphi(\omega) \rangle}_{L_x (\varphi(T_x \omega) - \varphi(\omega))} \right] = 0$$

$$= "L_\omega(x \cdot \varphi)".$$

$$= \mathbb{E} \left[ \sum_x w_{0,x} \underbrace{\varphi(x)}_{\varphi(-x)} \underbrace{\left( \varphi(\omega) - \varphi(T_{-x}(\omega)) \right)}_{-\nabla \varphi(\omega)} \right]$$

$$= -2 \mathbb{E} [\varphi L_\omega(x + \varphi)]$$

$L_\omega$  is a little different to  $L$ . Give  $\varphi$  is any function, you can do a shift on the environment.

$$T_y L_\omega(x + \varphi) = 0.$$

$$\overset{?}{\implies} \cancel{T_\omega(x + \varphi)} = 0 \quad \cancel{T_\omega(e_j \cdot x + \varphi_j(x))} = 0.$$

We will transfer this derivative about the environment to into  
need to

the derivative about the space. We will not expand on  
this.