

TCC: (Atypical) Behaviour
of random walks in
random or dynamic environment.

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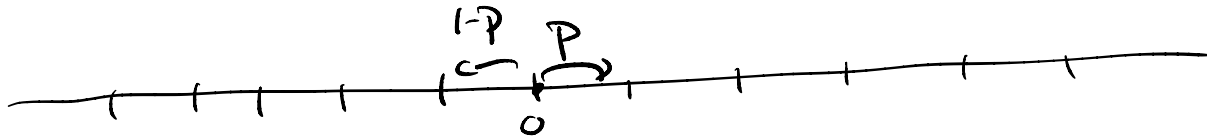
Notes on the webpage -

First, what's a typical ^{walk} with typical behaviour?

* Simple random walk on \mathbb{Z}

Markov Chain $(X_n)_{n \geq 0}$, $X_0 = 0$,
parameter $p \in (0, 1)$ **bias**

$$\mathbb{P}(X_{n+1} = \underset{-1}{x+1} \mid X_n = x) = \underset{1-p}{p}$$



* Law of large numbers

* If $P = \frac{1}{2}$, $\frac{X_n}{n} \rightarrow 0$ \mathbb{P} -a.s.

(X_n) will be recurrent
i.e. visit 0 infinitely many times.

* If $P \neq \frac{1}{2}$, $\frac{X_n}{n} \rightarrow v \neq 0$ | Ballistic
↑ speed (positive speed)

(X_n) is transient
i.e. visits 0 finitely many times.

* Central limit Theorem

Donsker's invariance principle

The law $\left(\frac{X_{\lfloor nt \rfloor} - nt\mu}{\sqrt{n}} \right)_{t \in [0,1]} \Rightarrow (B_t)_{t \in [0,1]}$

• $P = \frac{1}{2}$, $\frac{X_n}{\sqrt{n}} \rightarrow \mathcal{N}(0,1)$

Diffusive rescaling.

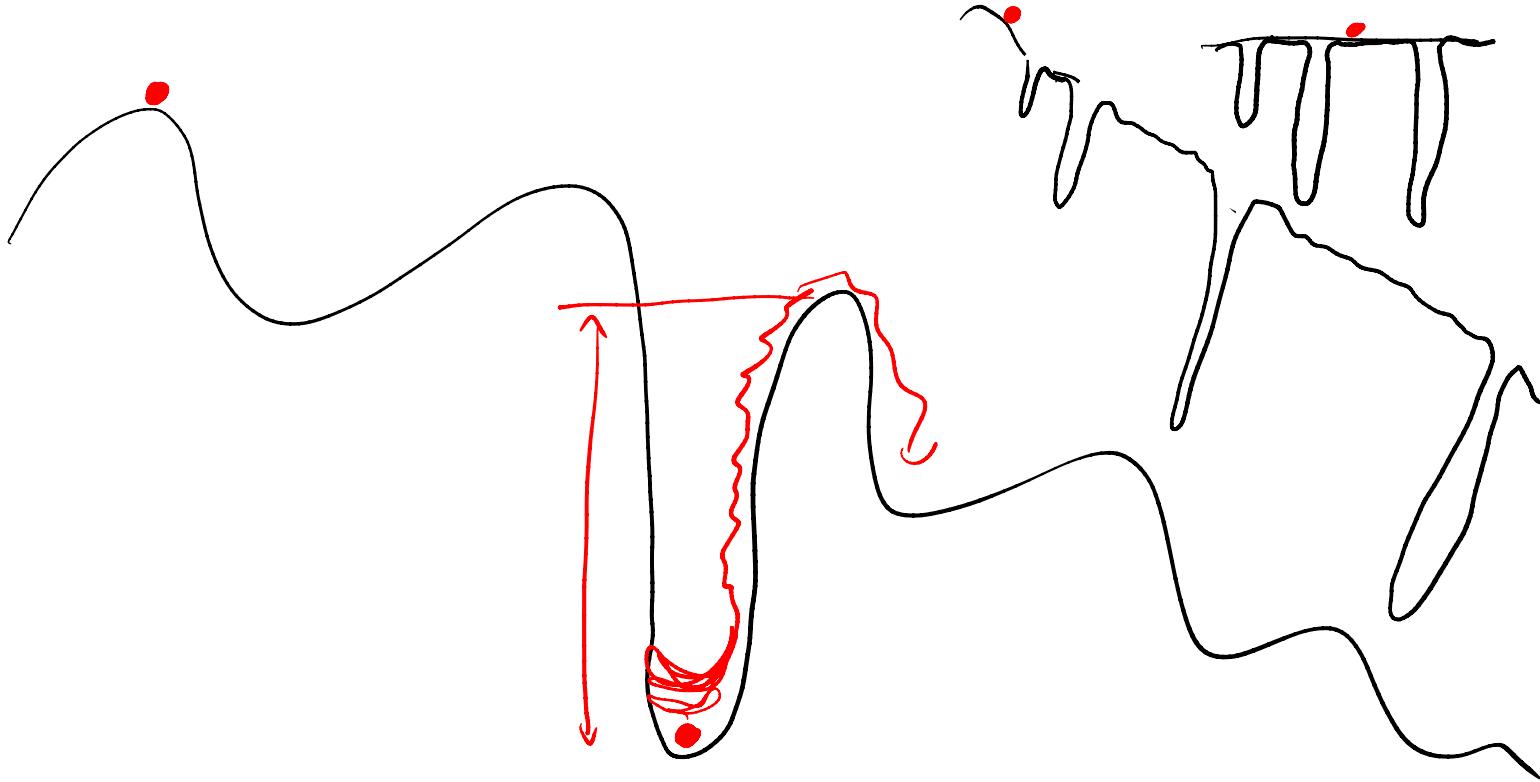
This is a typical behaviour

Either recurrent with $\mu=0$ or transient with $\mu \neq 0$

Typical position
at time n
is of order \sqrt{n} ,
after recentering.

What is atypical?

Random walk in random environment, on \mathbb{Z}

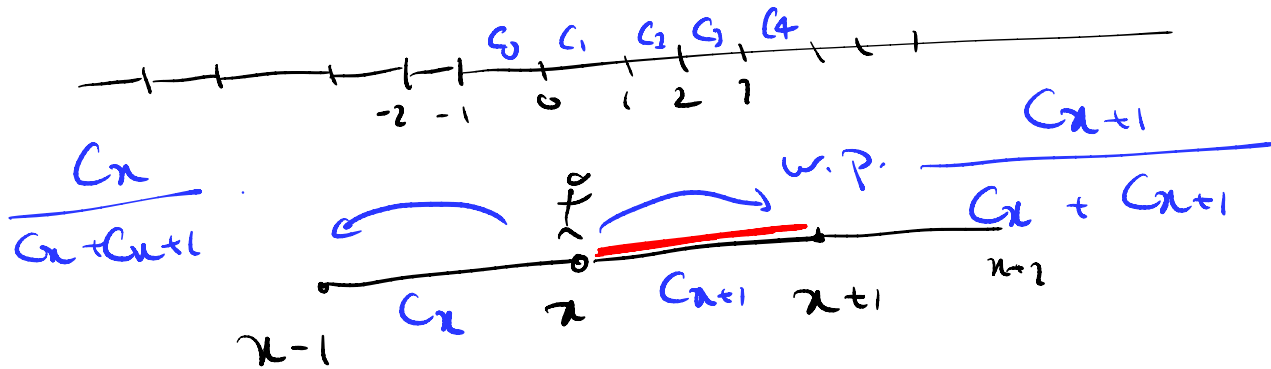


* How do we create these traps?

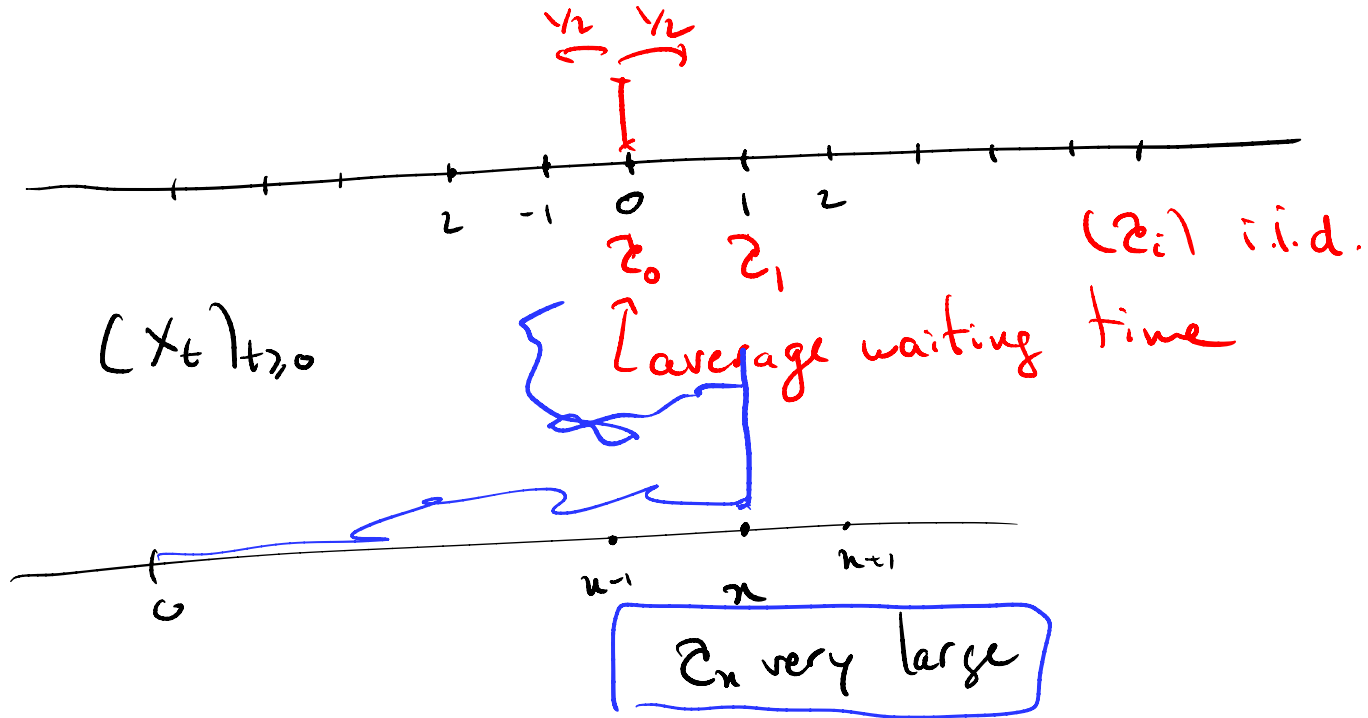
What are their size? their shape? Their effect?

* Can we have a characterization for the speed of a (trapped) random walk?

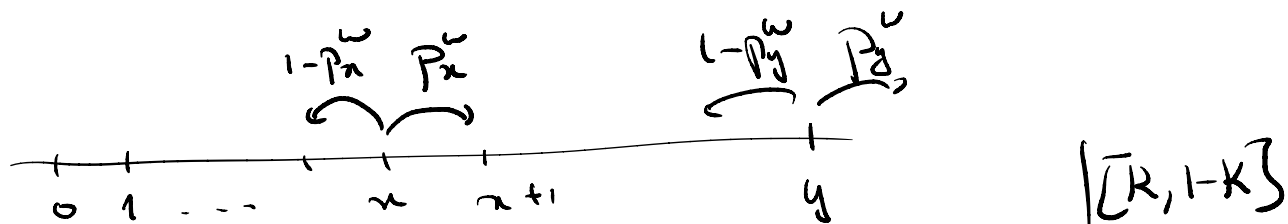
* Random walks on random conductances
 $c: \mathbb{C}(0, \infty)$



* Bouchaud trap model



* Random walk in random environment RWRE



$(\tilde{p}_n^w)_{n \in \mathbb{Z}}$ i.i.d in $[0, 1]$

on $d=1$: Problem | recurrence / transience
| ballisticity

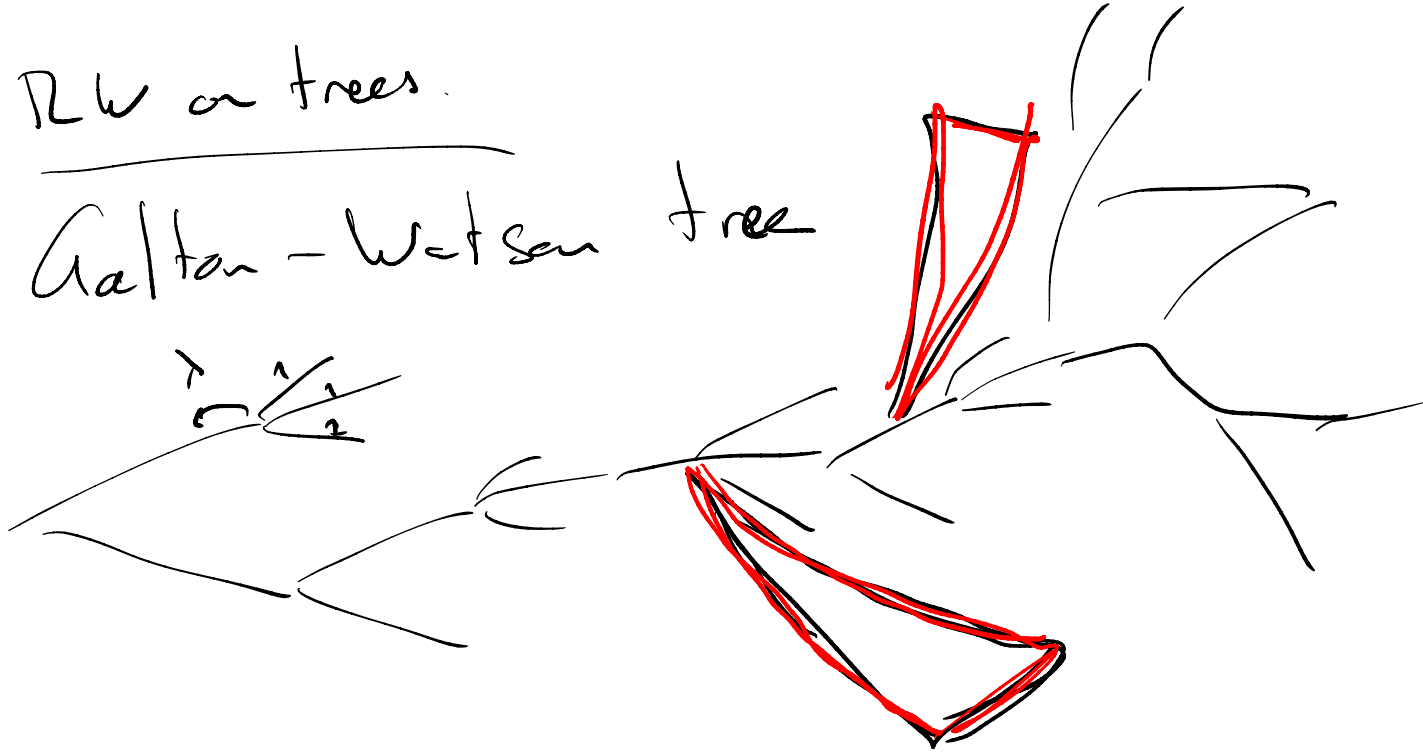
has been fully solved.

Explicit characterization.

on $d \geq 2$: Much more open
Squitman's condition [T].

RW on trees.

Galton-Watson tree



Consider a random walk on random conductances

Conductances: $C_n \in [K, +\infty)$ | ^{or} $[K, \frac{1}{K}]$
for some $K > 0$;

$\forall x \in \mathbb{Z}$.

Random walk $(X_n)_{n \geq 0}$ on the conductances (C_n)
is defined by $X_0 = 0$

$$P^w(X_{n+1} = \underset{-1}{x+1} \mid X_n = n) = \frac{C_{n+1}}{C_n + C_{n+1}} \\ = \frac{C_x}{C_n + C_{n+1}}$$

We will choose the C_n 's random,
i.i.d. over \mathbb{Z} , in $[K, +\infty)$ a.s.

\mathbb{P} : Their law.

If we fix the environment, i.e. collection $^w(C_n)$
and at the evolution of X

\mathbb{P}_0^w = quenched law of X

If we average over the environment:

$\mathbb{P}_0 = \mathbb{E}[\mathbb{P}_0^w(\cdot)]$ annealed law

under \mathbb{P}_0 , X is not a Markov Chain

Crash course on RW on networks

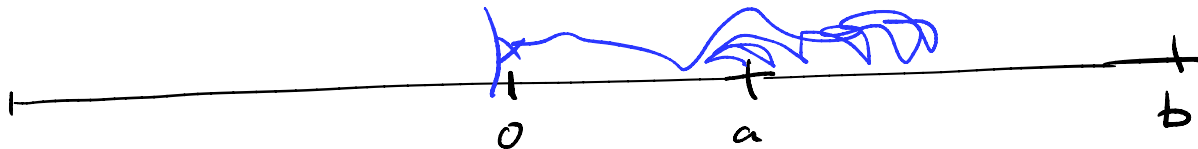
edge c_e : conductances
 $r_e = \frac{1}{c_e}$: resistances

* Electrical network.

Effective resistance between two points a & z
Conductance

$$C_{\text{eff}}(a \leftrightarrow z) = \pi(a) \mathbb{P}(a \rightarrow z) = \frac{1}{R_{\text{eff}}(a \leftrightarrow z)}$$

$\left[\sum_{y \sim a} C(a, y) \right]$



$$\mathbb{P}_a(T_0 < T_b) = \frac{R_{\text{eff}}(a \leftrightarrow b)}{R_{\text{eff}}(0 \leftrightarrow b)}$$

↖ hitting times

$$R_{\text{eff}}(0 \leftrightarrow b) = R_{\text{eff}}(0 \leftrightarrow a) + R_{\text{eff}}(a \leftrightarrow b)$$

Back on \mathbb{Z} : RW X on (\mathbb{C}_n) is reversible under P_0^μ .
 (left as an exercise)

measure $\pi(n) = C_n + C_{n+1}$ is

reversible.

Also true for RWRC on \mathbb{Z}^d , $d \geq 1$.

* Assume $C_n \in [K, \frac{1}{K}]$ \mathbb{P} -a.s. for some $K \in (0, 1)$.

"Easy" case.

We can prove $\frac{X_n}{n} \rightarrow 0$ \mathbb{P} -a.s.

and we can prove diffusive fluctuations

Let's prove these results using the environment
seen from the particle
walker

shift
along trajectories

$$\bar{\omega}_n = \tau_{X_n} \omega$$

$$\bar{\omega}_1 \quad \bar{\omega}_0 \quad \bar{\omega}_1 \quad \bar{\omega}_2$$



$\bar{\omega}_0 = \omega$, $(\bar{\omega}_n)_n$ is a Markov process on the space of the environment

\leadsto \exists a measure \mathbb{Q} ergodic, invariant, probability measure.

Under \mathbb{Q} , the law of $\bar{\omega}_1$ is \mathbb{Q}

$$\mathbb{Q} \sim \mathbb{P}$$

$$\mathbb{Q} = \frac{1}{Z} (G + A) \mathbb{P}, \quad Z \text{ normalising Cst.}$$

Consider the martingale

$$\begin{aligned} \Pi_n &= X_n - \sum_{k=0}^{n-1} E_{X_k}^{\omega} [X_1] = E[X_{n+1} - X_n | \mathcal{F}_n] \\ &= X_n - \sum_{k=0}^{n-1} d(X_k, \omega) = \frac{C_{1+X_k} - C_{X_k}}{C_{X_k} + C_{1+X_k}} \end{aligned}$$

local drift at X_n .

Exercise: Prove $E[\Pi_{n+1} - \Pi_n | \mathcal{F}_n] = 0$,
 $\mathcal{F}_n = \sigma(X_0, \dots, X_n)$.

Azuma's inequality \Rightarrow almost surely,

$$\Pi_n \leq n^{\frac{1}{2} + \varepsilon} \text{ for } n \text{ large enough.}$$

$$\Rightarrow \frac{\Pi_n}{n} \rightarrow 0 \quad \mathbb{P}_0^{\omega}\text{-a.e. for } \mathbb{P}\text{-a.e. } \omega$$

$$0 \stackrel{\text{a.s.}}{<} \frac{\bar{X}_n}{n} = \frac{X_n}{n} - \underbrace{\frac{1}{n} \sum_{k=0}^{n-1} d(X_k, \omega)}_{\rightarrow 0}$$

$$\frac{1}{n} \sum_{k=0}^{n-1} d(X_k, \omega) = \frac{1}{n} \sum_{k=0}^{n-1} d(0, \bar{\omega}_k)$$

We will use the Birkhoff's Ergodic Theorem

(Y, \mathcal{B}, T, μ) , f : measurable

$$E_T[|f|] < \infty$$

T : measure preserving transformation

$$\mu(T^{-1}(A)) = \mu(A)$$

Then w.p. 1

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} f(T^k \omega) = \underbrace{E_T[f | \mathcal{C}]}(\omega)$$

$$\left\{ \begin{array}{l} \mathcal{C} : \sigma\text{-field generated by} \\ \text{invariant sets of } T \\ \mu(T^{-1}(E) \Delta E) = 0 \end{array} \right.$$

Pointwise E.T. || If T is ergodic (i.e. E invariant $\Rightarrow \mu(E) \in \{0, 1\}$)
 then \mathcal{C} is trivial

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} f(T^k x) = \int f d\mu = \mathbb{E}_\mu[f](x).$$

This implies, for \mathbb{Q} -a.e. ω

$$\frac{1}{n} \sum_{k=0}^{n-1} d(o, \bar{w}_k) \xrightarrow{P_o^\omega} \mathbb{Q}[d(o, \omega)]$$

This also holds for \mathbb{P} -a.e. ω . ($\mathbb{Q} \sim \mathbb{P}$).

$$\frac{\Pi_n}{n} \rightarrow 0 \text{ a.s.}$$

we have that for P-a.e. ω ,

$$\frac{X_n}{n} \xrightarrow{P_0^{\omega} \text{ a.s.}} Q[d(\varnothing, \omega)] = \underline{0}$$

Fluctuations? Use a martingale CLT

Martingale CLT: (Π_n) an (\mathcal{F}_n) -martingale, square

integrable: for each $n \in \mathbb{N}$, $E[|\Pi_n|^2] < +\infty$.

II
(i) $\exists \sigma^2 \in [0, \infty)$ s.t., $\forall t > 0$,

$$\frac{1}{n} \sum_{k=0}^{Ltn} E[|\Pi_{k+1} - \Pi_k|^2 | \mathcal{F}_k] \rightarrow t \cdot \sigma^2 \text{ in probability.}$$

(2) $\forall \varepsilon > 0$

$$\frac{1}{n} \sum_{k=0}^{n-1} \mathbb{E} \left[|\Pi_{k+1} - \Pi_k|^2 \mathbb{1}_{\{|\Pi_{k+1} - \Pi_k| > \varepsilon \sqrt{n}\}} \right] \xrightarrow{\mathbb{P}} 0$$



Then the law of $\left(\frac{\Pi_{\lfloor nt \rfloor}}{\sqrt{n}} \right)_{t \in [0, T]}$ $\xrightarrow{(d)}$ $(B_t)_{t \in [0, T]}$

interpolated

variance σ^2 .

If $|\Pi_{k+1} - \Pi_k| < C\sqrt{n}$ then (2) is satisfied.

For us, let consider

$$\tilde{\Pi}_n = \begin{cases} R(0, X_n) & \text{if } X_n \geq 0 \\ -R(0, X_n) & \text{if } X_n < 0 \end{cases}$$

$$\begin{cases} R(0, x) \\ = \sum_{i=1}^n r_i \\ \frac{1}{c_i} \end{cases}$$

$(\tilde{\pi}_n)$ is a martingale (left as exercise)

to apply the mart. CLT, we need to prove

2) $(\tilde{\pi}_n)$ has bounded steps:

true because $r_i < \frac{1}{K} \forall i \in \mathbb{Z}$.

$$1) \frac{1}{n} \sum_{k=0}^{\lfloor tn \rfloor} \mathbb{E} \left[|\tilde{\pi}_{n+1} - \tilde{\pi}_n|^2 \mid \mathcal{F}_n \right] \neq ?$$

$$\frac{C_{X_{n+1}}}{C_{X_n} + C_{X_{n+1}}} \cdot r_{X_{n+1}}^2 + \frac{C_{X_n}}{C_{X_n} + C_{1+X_n}} \cdot r_{X_n}^2$$

$$= \frac{r_{X_n} r_{1+X_n}^2}{r_{X_n} + r_{1+X_n}} + \frac{r_{1+X_n} r_{X_n}^2}{r_{X_n} + r_{1+X_n}} = r_{X_n} r_{X_{n+1}}$$

→ Ergodic Theorem

$$\rightarrow t_x \frac{1}{nt} \times \sum_{k=0}^{[nt]} E^x [| \Pi_{kx} - \Pi_n |^2 | \mathcal{F}_n]$$

$$\xrightarrow{P_0^u} t_x Q(r_0, r_1) \quad \text{for } \begin{cases} Q\text{-a.e. w} \\ P\text{-a.e.} \end{cases}$$

$$\Rightarrow \left(\frac{R(0, X_{[nt]})}{\sqrt{n}} \right) \Rightarrow (B_t)_{t \in \mathbb{R}}$$

$$\Rightarrow \left(\frac{X_{[nt]}}{\sqrt{n}} \right) \Rightarrow (B_t) \quad \text{because } \checkmark \text{ n large.}$$

$R(0, x) \sim x \cdot E[r_1]$

Next week: What happens
if C_n can be very large??