

Exponential concentration of cover times

Alex Zhai (azhai@stanford.edu)

June 23, 2016

- Part I: Preliminaries
 - Effective resistance and Gaussian free fields
 - Ray-Knight theorems
- Part II: Application to cover times
- Part III: Stochastic domination in the generalized 2nd Ray-Knight theorem

Part I: Preliminaries

- $G = (V, E)$ a simple graph, and fix a starting vertex $v_0 \in V$.

Our setting

- $G = (V, E)$ a simple graph, and fix a starting vertex $v_0 \in V$.
- We consider **continuous time** random walks $X = \{X_t\}_{t \in \mathbb{R}^+}$ started at v_0 :
 - same as usual simple random walk, except time between jumps is a standard exponential random variable
 - X_t denotes the vertex you're on at time t

- $G = (V, E)$ a simple graph, and fix a starting vertex $v_0 \in V$.
- We consider **continuous time** random walks $X = \{X_t\}_{t \in \mathbb{R}^+}$ started at v_0 :
 - same as usual simple random walk, except time between jumps is a standard exponential random variable
 - X_t denotes the vertex you're on at time t
- Define
 - **cover time**

τ_{cov} = the first time all vertices are visited at least once

- **hitting time**

$\tau_{\text{hit}}(x, y)$ = the first time walk started at x visits y

Effective resistance

- For any $x, y \in V$, imagine all the edges are unit resistors and we connect the ends of a battery to x and y .

Effective resistance

- For any $x, y \in V$, imagine all the edges are unit resistors and we connect the ends of a battery to x and y . Then, define

$$R_{\text{eff}}(x, y) = \text{effective resistance between } x \text{ and } y$$

Effective resistance

- For any $x, y \in V$, imagine all the edges are unit resistors and we connect the ends of a battery to x and y . Then, define

$$R_{\text{eff}}(x, y) = \text{effective resistance between } x \text{ and } y$$

- We can compute $R_{\text{eff}}(x, y)$ by solving for a function $f : V \rightarrow \mathbb{R}$ such that

$$\Delta f(z) = \begin{cases} 1 & \text{if } z = x \\ -1 & \text{if } z = y \\ 0 & \text{otherwise} \end{cases}$$

Then $R_{\text{eff}}(x, y) = f(y) - f(x)$.

- For any $x, y \in V$, imagine all the edges are unit resistors and we connect the ends of a battery to x and y . Then, define

$$R_{\text{eff}}(x, y) = \text{effective resistance between } x \text{ and } y$$

- We can compute $R_{\text{eff}}(x, y)$ by solving for a function $f : V \rightarrow \mathbb{R}$ such that

$$\Delta f(z) = \begin{cases} 1 & \text{if } z = x \\ -1 & \text{if } z = y \\ 0 & \text{otherwise} \end{cases}$$

Then $R_{\text{eff}}(x, y) = f(y) - f(x)$.

- Commutate time identity:

$$\frac{\mathbf{E}\tau_{\text{hit}}(x, y) + \mathbf{E}\tau_{\text{hit}}(y, x)}{2} = |E| \cdot R_{\text{eff}}(x, y).$$

Gaussian free field: definition

Gaussian free field: definition

For a graph $G = (V, E)$, the **Gaussian free field** (GFF) η is a multivariate Gaussian:

- coordinates η_v indexed by $v \in V$, with $\eta_{v_0} = 0$

Gaussian free field: definition

For a graph $G = (V, E)$, the **Gaussian free field** (GFF) η is a multivariate Gaussian:

- coordinates η_v indexed by $v \in V$, with $\eta_{v_0} = 0$
- for $f \in \mathbb{R}^V$ with $f_{v_0} = 0$,

$$[\text{probability of } f] \propto \exp\left(-\frac{1}{2} \sum_{(x,y) \in E} (f_x - f_y)^2\right)$$

Gaussian free field: definition

For a graph $G = (V, E)$, the **Gaussian free field** (GFF) η is a multivariate Gaussian:

- coordinates η_v indexed by $v \in V$, with $\eta_{v_0} = 0$
- for $f \in \mathbb{R}^V$ with $f_{v_0} = 0$,

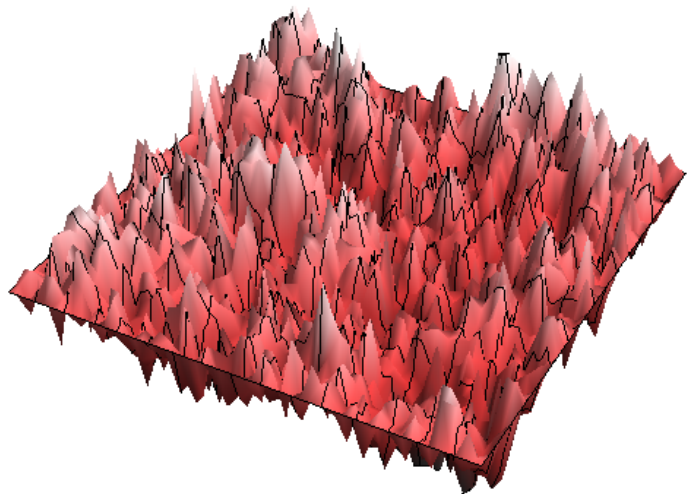
$$[\text{probability of } f] \propto \exp\left(-\frac{1}{2} \sum_{(x,y) \in E} (f_x - f_y)^2\right)$$

- equivalently,

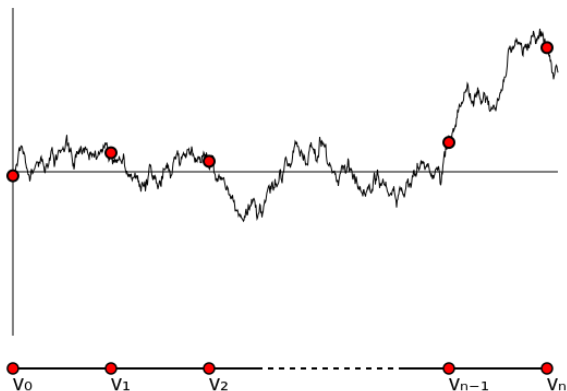
$$\mathbf{E}(\eta_x - \eta_y)^2 = R_{\text{eff}}(x, y) \quad (\text{note: } \mathbf{E}\eta_x^2 = R_{\text{eff}}(x, v_0))$$

Gaussian free field: example

Below is a realization of the GFF on a discrete 2D lattice:



Gaussian free field: example



Let $\{B_t\}_{t \geq 0}$ be a Brownian motion. GFF of a path is

$$\eta = (0 = B_0, B_1, \dots, B_n).$$

- Reminder: $G = (V, E)$ a graph and X_t a continuous time random walk.

- Reminder: $G = (V, E)$ a graph and X_t a continuous time random walk.
- For $x \in V$ and $s \in \mathbb{R}^+$, define **local time**

$$\begin{aligned}\mathcal{L}_s(x) &= \frac{1}{\deg(x)} \int_0^s \mathbf{1}(X_{s'} = x) ds' \\ &= \frac{1}{\deg(x)} (\text{time spent by r.w. at } x \text{ up to time } s).\end{aligned}$$

- For any $t > 0$, define

$$\begin{aligned}\tau^+(t) &= \inf\{s \geq 0 : \mathcal{L}_s(v_0) \geq t\} \\ &= \text{first time that } v_0 \text{ accumulates local time } t.\end{aligned}$$

- For any $t > 0$, define

$$\begin{aligned}\tau^+(t) &= \inf\{s \geq 0 : \mathcal{L}_s(v_0) \geq t\} \\ &= \text{first time that } v_0 \text{ accumulates local time } t.\end{aligned}$$

- Remark: $\tau^+\left(\frac{1}{\deg(v_0)}\right)$ is like the return time of a discrete time random walk.

- For any $t > 0$, define

$$\begin{aligned}\tau^+(t) &= \inf\{s \geq 0 : \mathcal{L}_s(v_0) \geq t\} \\ &= \text{first time that } v_0 \text{ accumulates local time } t.\end{aligned}$$

- Remark: $\tau^+\left(\frac{1}{\deg(v_0)}\right)$ is like the return time of a discrete time random walk.
- We have

$$\mathbf{E}\tau^+(t) = 2|E| \cdot t.$$

(Analogous to expected return time being equal to inverse stationary probability.)

Generalized 2nd Ray-Knight theorem

Theorem (Generalized Second Ray-Knight Theorem)

Let X be a continuous time random walk, and let η and η' be GFFs with X and η independent. Then, for any $t > 0$,

$$\left\{ \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} \stackrel{\text{law}}{=} \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}.$$

Generalized 2nd Ray-Knight theorem

Theorem (Generalized Second Ray-Knight Theorem)

Let X be a continuous time random walk, and let η and η' be GFFs with X and η independent. Then, for any $t > 0$,

$$\left\{ \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} \stackrel{\text{law}}{=} \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}.$$

Above theorem due to Eisenbaum-Kaspi-Marcus-Rosen-Shi.

Generalized 2nd Ray-Knight theorem

Theorem (Generalized Second Ray-Knight Theorem)

Let X be a continuous time random walk, and let η and η' be GFFs with X and η independent. Then, for any $t > 0$,

$$\left\{ \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2 \right\}_{x \in V} \stackrel{\text{law}}{=} \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}.$$

Above theorem due to Eisenbaum-Kaspi-Marcus-Rosen-Shi.

Similar/related theorems by Ray, Knight, Dynkin, Le Jan, Sznitman, and others.

Part II: Application to cover times

Theorem (Borell and Sudakov-Tsirelson)

Let $\eta = \{\eta_i\}_{i \in I}$ be any centered multivariate Gaussian with $\mathbf{E}\eta_i^2 \leq \sigma^2$ for each i . Let

$$X = \sup_{i \in I} \eta_i.$$

Then,

$$\mathbf{P}(|X - \mathbf{E}X| > s \cdot \sigma) \leq 2(1 - \Phi(s)),$$

where Φ is the Gaussian CDF.

In other words, the maximum (or minimum) of a Gaussian process is at least as concentrated as a Gaussian.

- Define

$$R = \max_{x,y \in V} R_{\text{eff}}(x,y) \geq \max_{x \in V} \mathbf{E} \eta_x^2$$

$$M = \mathbf{E} \max_{v \in V} \eta_v = -\mathbf{E} \min_{v \in V} \eta_v.$$

- Define

$$R = \max_{x,y \in V} R_{\text{eff}}(x,y) \geq \max_{x \in V} \mathbf{E} \eta_x^2$$

$$M = \mathbf{E} \max_{v \in V} \eta_v = -\mathbf{E} \min_{v \in V} \eta_v.$$

- By Gaussian concentration, $\max_{v \in V} \eta_v$ has mean M and fluctuations of order \sqrt{R} .

- Define

$$R = \max_{x,y \in V} R_{\text{eff}}(x,y) \geq \max_{x \in V} \mathbf{E} \eta_x^2$$

$$M = \mathbf{E} \max_{v \in V} \eta_v = -\mathbf{E} \min_{v \in V} \eta_v.$$

- By Gaussian concentration, $\max_{v \in V} \eta_v$ has mean M and fluctuations of order \sqrt{R} .
- In many cases, $\sqrt{R} \ll M$.
 - e.g. complete graph, discrete torus, regular trees
 - **doesn't** hold for case of a path

Theorem (generalized Ray-Knight)

$$\left\{ \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} \stackrel{\text{law}}{=} \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}$$

Theorem (generalized Ray-Knight)

$$\left\{ \mathcal{L}_{\tau^+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} \stackrel{\text{law}}{=} \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}$$

Main observation (Ding-Lee-Peres):

$$\tau^+(t) < \tau_{\text{cov}} \iff \text{one of the } \mathcal{L}_{\tau^+(t)}(x) \text{ is } 0$$

Theorem (generalized Ray-Knight)

$$\left\{ \mathcal{L}_{\tau^+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} \stackrel{\text{law}}{=} \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}$$

Main observation (Ding-Lee-Peres):

$$2|E| \cdot t \approx \tau^+(t) < \tau_{\text{cov}} \quad \iff \quad \text{one of the } \mathcal{L}_{\tau^+(t)}(x) \text{ is } 0$$

Theorem (generalized Ray-Knight)

$$\left\{ \mathcal{L}_{\tau^+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} \stackrel{\text{law}}{=} \left\{ \frac{1}{2} (\eta'_x + \sqrt{2t})^2 \right\}_{x \in V}$$

Main observation (Ding-Lee-Peres):

$$2|E| \cdot t \approx \tau^+(t) < \tau_{\text{cov}} \quad \begin{array}{l} \iff \\ \text{“} \iff \text{”} \end{array} \quad \begin{array}{l} \text{one of the } \mathcal{L}_{\tau^+(t)}(x) \text{ is } 0 \\ \text{one of the } (\eta'_x + \sqrt{2t}) \text{ is small} \end{array}$$

Theorem (generalized Ray-Knight)

$$\left\{ \mathcal{L}_{\tau^+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} \stackrel{\text{law}}{=} \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}$$

Main observation (Ding-Lee-Peres):

$$\begin{aligned} 2|E| \cdot t \approx \tau^+(t) < \tau_{\text{cov}} & \iff \text{one of the } \mathcal{L}_{\tau^+(t)}(x) \text{ is } 0 \\ \text{"} \iff \text{"} & \iff \text{one of the } (\eta'_x + \sqrt{2t}) \text{ is small} \\ \text{"} \iff \text{"} & \iff \min_{x \in V} \eta'_x < -\sqrt{2t} \end{aligned}$$

Theorem (generalized Ray-Knight)

$$\left\{ \mathcal{L}_{\tau^+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} \stackrel{\text{law}}{=} \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}$$

Main observation (Ding-Lee-Peres):

$$\begin{aligned} 2|E| \cdot t \approx \tau^+(t) < \tau_{\text{cov}} & \iff \text{one of the } \mathcal{L}_{\tau^+(t)}(x) \text{ is } 0 \\ \text{"} \iff \text{"} & \iff \text{one of the } (\eta'_x + \sqrt{2t}) \text{ is small} \\ \text{"} \iff \text{"} & \iff \min_{x \in V} \eta'_x < -\sqrt{2t} \end{aligned}$$

Critical value of t occurs around $\sqrt{2t} = M$.

Theorem (generalized Ray-Knight)

$$\left\{ \mathcal{L}_{\tau^+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} \stackrel{\text{law}}{=} \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}$$

Main observation (Ding-Lee-Peres):

$$\begin{aligned} 2|E| \cdot t \approx \tau^+(t) < \tau_{\text{cov}} & \iff \text{one of the } \mathcal{L}_{\tau^+(t)}(x) \text{ is } 0 \\ \text{“} \iff \text{”} & \iff \text{one of the } (\eta'_x + \sqrt{2t}) \text{ is small} \\ \text{“} \iff \text{”} & \iff \min_{x \in V} \eta'_x < -\sqrt{2t} \end{aligned}$$

Critical value of t occurs around $\sqrt{2t} = M$.

Theorem (Ding-Lee-Peres)

$$\mathbf{E} \tau_{\text{cov}} \asymp |E| \cdot \left(-\mathbf{E} \min_{x \in V} \eta'_x \right)^2 = |E| \cdot M^2.$$

Statement of the concentration bound

Theorem (Z., following conjecture of Ding)

There are universal constants c and C such that

$$\mathbf{P} \left(\left| \tau_{\text{cov}} - |E|M^2 \right| \geq |E|(\sqrt{\lambda R} \cdot M + \lambda R) \right) \leq Ce^{-c\lambda}$$

for any $\lambda \geq C$.

Statement of the concentration bound

Theorem (Z., following conjecture of Ding)

There are universal constants c and C such that

$$\mathbf{P} \left(\left| \tau_{\text{cov}} - |E|M^2 \right| \geq |E|(\sqrt{\lambda R} \cdot M + \lambda R) \right) \leq Ce^{-c\lambda}$$

for any $\lambda \geq C$.

Recall:

$$\max_{x,y \in V} \mathbf{E} \tau_{\text{hit}}(x,y) \asymp |E| \cdot R \quad \text{and} \quad \mathbf{E} \tau_{\text{cov}} \asymp |E| \cdot M^2.$$

Statement of the concentration bound

Theorem (Z., following conjecture of Ding)

There are universal constants c and C such that

$$\mathbf{P} \left(\left| \tau_{\text{cov}} - |E|M^2 \right| \geq |E|(\sqrt{\lambda R} \cdot M + \lambda R) \right) \leq Ce^{-c\lambda}$$

for any $\lambda \geq C$.

Recall:

$$\max_{x,y \in V} \mathbf{E} \tau_{\text{hit}}(x,y) \asymp |E| \cdot R \quad \text{and} \quad \mathbf{E} \tau_{\text{cov}} \asymp |E| \cdot M^2.$$

Thus,

$$\mathbf{E} \tau_{\text{cov}} \sim |E| \cdot M^2 \quad \text{whenever} \quad \max_{x,y \in V} \mathbf{E} \tau_{\text{hit}}(x,y) \ll \mathbf{E} \tau_{\text{cov}}.$$

Statement of the concentration bound

Theorem (Z., following conjecture of Ding)

There are universal constants c and C such that

$$\mathbf{P} \left(\left| \tau_{\text{cov}} - |E|M^2 \right| \geq |E|(\sqrt{\lambda R} \cdot M + \lambda R) \right) \leq Ce^{-c\lambda}$$

for any $\lambda \geq C$.

Recall:

$$\max_{x,y \in V} \mathbf{E} \tau_{\text{hit}}(x,y) \asymp |E| \cdot R \quad \text{and} \quad \mathbf{E} \tau_{\text{cov}} \asymp |E| \cdot M^2.$$

Thus,

$$\mathbf{E} \tau_{\text{cov}} \sim |E| \cdot M^2 \quad \text{whenever} \quad \max_{x,y \in V} \mathbf{E} \tau_{\text{hit}}(x,y) \ll \mathbf{E} \tau_{\text{cov}}.$$

(Ding proved for trees and bounded degree graphs.)

Upper bound (following Ding-Lee-Peres)

$$A_x = \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2, \quad B_x = \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2.$$

Suppose $\mathbf{P}(\mathcal{L}_{\tau+(t)}(x) = 0$ for some x) is large.

Upper bound (following Ding-Lee-Peres)

$$A_x = \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2, \quad B_x = \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2.$$

Suppose $\mathbf{P}(\mathcal{L}_{\tau+(t)}(x) = 0 \text{ for some } x)$ is large. Then,

$$\mathbf{P}\left(\min_{x \in V} A_x < R\right) \geq \mathbf{P}(\mathcal{L}_{\tau+(t)}(x) = 0 \text{ for some } x) \cdot \underbrace{\mathbf{P}(\eta_x^2 < R)}_{\geq 0.5}$$

is large,

Upper bound (following Ding-Lee-Peres)

$$A_x = \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2, \quad B_x = \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2.$$

Suppose $\mathbf{P}(\mathcal{L}_{\tau+(t)}(x) = 0 \text{ for some } x)$ is large. Then,

$$\mathbf{P} \left(\min_{x \in V} A_x < R \right) \geq \mathbf{P}(\mathcal{L}_{\tau+(t)}(x) = 0 \text{ for some } x) \cdot \underbrace{\mathbf{P}(\eta_x^2 < R)}_{\geq 0.5}$$

is large, so

$$\mathbf{P} \left(\min_{x \in V} B_x < R \right) = \mathbf{P} \left(\min_{x \in V} A_x < R \right)$$

is large,

Upper bound (following Ding-Lee-Peres)

$$A_x = \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2, \quad B_x = \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2.$$

Suppose $\mathbf{P}(\mathcal{L}_{\tau+(t)}(x) = 0 \text{ for some } x)$ is large. Then,

$$\mathbf{P} \left(\min_{x \in V} A_x < R \right) \geq \mathbf{P}(\mathcal{L}_{\tau+(t)}(x) = 0 \text{ for some } x) \cdot \underbrace{\mathbf{P}(\eta_x^2 < R)}_{\geq 0.5}$$

is large, so

$$\mathbf{P} \left(\min_{x \in V} B_x < R \right) = \mathbf{P} \left(\min_{x \in V} A_x < R \right)$$

is large, which means $\sqrt{2t}$ can't be much more than M .

Lower bound (following Ding)

$$A_x = \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2, \quad B_x = \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2.$$

Suppose $\sqrt{2t} < M - C\sqrt{R}$.

Lower bound (following Ding)

$$A_x = \mathcal{L}_{\tau^+(t)}(x) + \frac{1}{2}\eta_x^2, \quad B_x = \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2.$$

Suppose $\sqrt{2t} < M - C\sqrt{R}$. Then

$$\mathbf{P} \left(\min_{x \in V} \eta'_x + \sqrt{2t} < 0 \right) = \mathbf{P} \left(\min_{x \in V} \eta'_x < -M + C\sqrt{R} \right)$$

is large (for C large, think e.g. $C = 10$)...

Lower bound (following Ding)

$$A_x = \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2, \quad B_x = \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2.$$

Suppose $\sqrt{2t} < M - C\sqrt{R}$. Then

$$\mathbf{P} \left(\min_{x \in V} \eta'_x + \sqrt{2t} < 0 \right) = \mathbf{P} \left(\min_{x \in V} \eta'_x < -M + C\sqrt{R} \right)$$

is large (for C large, think e.g. $C = 10$)... and

$$\eta'_x + \sqrt{2t} < 0 \text{ for some } x \quad \text{“} \implies \text{”} \quad \underline{\mathcal{L}_{\tau+(t)}(x) = 0}.$$

Lower bound (following Ding)

$$A_x = \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2, \quad B_x = \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2.$$

Suppose $\sqrt{2t} < M - C\sqrt{R}$. Then

$$\mathbf{P} \left(\min_{x \in V} \eta'_x + \sqrt{2t} < 0 \right) = \mathbf{P} \left(\min_{x \in V} \eta'_x < -M + C\sqrt{R} \right)$$

is large (for C large, think e.g. $C = 10$)... and

$$\eta'_x + \sqrt{2t} < 0 \text{ for some } x \quad \text{“} \implies \text{”} \quad \underline{\mathcal{L}_{\tau+(t)}(x) = 0}.$$

Important missing step: how to make “ \implies ” rigorous.

- The transition point of whether

$$\mathcal{L}_{\tau+(t)}(x) > 0 \text{ for all } x \in V$$

occurs around $\sqrt{2t} \approx M \implies t \approx \frac{1}{2}M^2$.

Concentration of cover times: recap

- The transition point of whether

$$\mathcal{L}_{\tau^+(t)}(x) > 0 \text{ for all } x \in V$$

occurs around $\sqrt{2t} \approx M \implies t \approx \frac{1}{2}M^2$.

- $\tau^+(t)$ is concentrated around its expectation $2|E| \cdot t$ as long as $R \ll t$, so

$$\tau_{\text{cov}} \approx \tau^+ \left(\frac{1}{2}M^2 \right) \approx |E| \cdot M^2.$$

- The transition point of whether

$$\mathcal{L}_{\tau^+(t)}(x) > 0 \text{ for all } x \in V$$

occurs around $\sqrt{2t} \approx M \implies t \approx \frac{1}{2}M^2$.

- $\tau^+(t)$ is concentrated around its expectation $2|E| \cdot t$ as long as $R \ll t$, so

$$\tau_{\text{cov}} \approx \tau^+ \left(\frac{1}{2}M^2 \right) \approx |E| \cdot M^2.$$

- But still need “important missing step”.

Part III: Stochastic domination in the generalized 2nd Ray-Knight theorem

Theorem (variant of theorem of Lupu, conjectured by Ding)

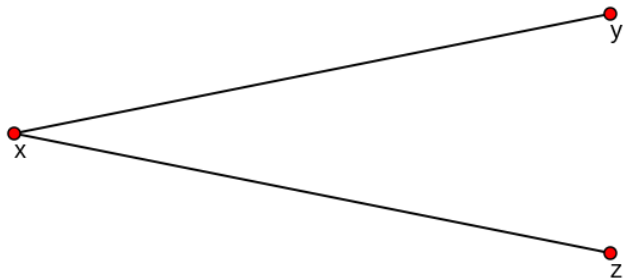
We have

$$\left\{ \sqrt{\mathcal{L}_{\tau+(t)}(x)} : x \in V \right\} \preceq \frac{1}{\sqrt{2}} \left\{ \max(\eta'_x + \sqrt{2t}, 0) : x \in V \right\},$$

where \preceq denotes stochastic domination.

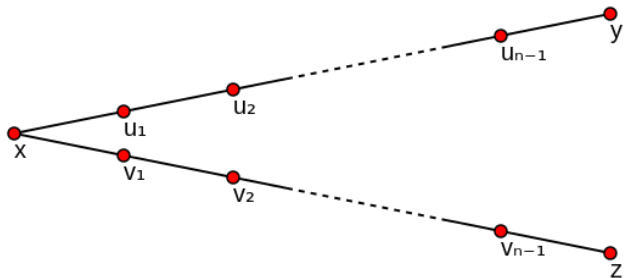
A graph refinement

Random walk step can be simulated by random walk on refined graph:



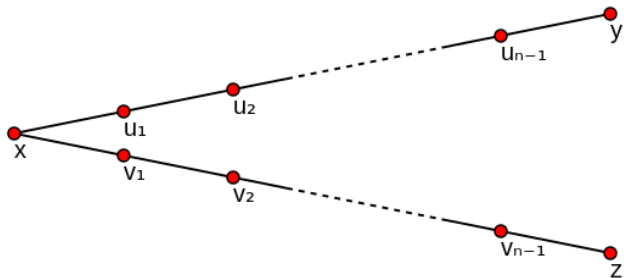
A graph refinement

Random walk step can be simulated by random walk on refined graph:



A graph refinement

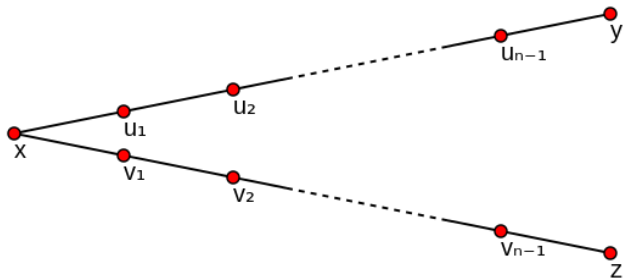
Random walk step can be simulated by random walk on refined graph:



Refined walk visits x a **Geom**(n) number of times before going to y or z with equal probability

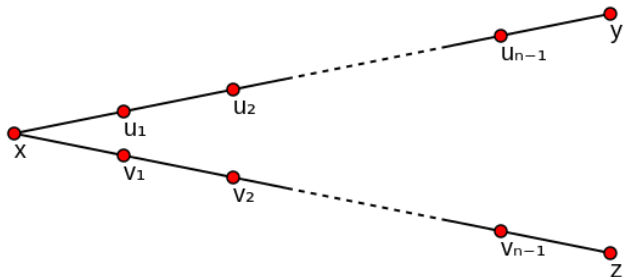
A graph refinement

Random walk step can be simulated by random walk on refined graph:



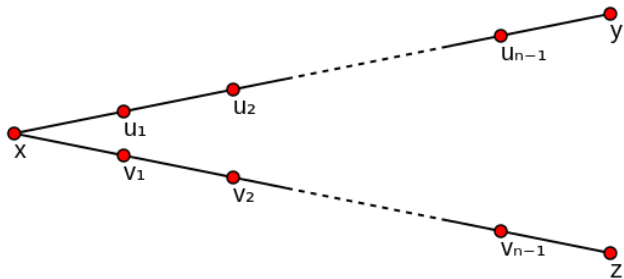
Refined walk visits x a **Geom**(n) number of times before going to y or z with equal probability \implies time spent at x is still exponential

A graph refinement



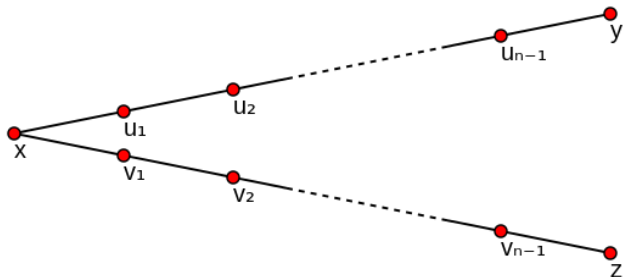
The GFFs are also related in a natural way: effective resistances (= GFF covariances) are multiplied by n .

Metric graphs



The limiting object as $n \rightarrow \infty$ is known as a **metric graph**.

Metric graphs



The limiting object as $n \rightarrow \infty$ is known as a **metric graph**. In the limit:

- random walk is a “Brownian motion on edges”.
- GFF has same law as original graph (up to scaling), with Brownian bridges on edges

Artificially construct a coupling of random walk X and GFFs η and η' on the metric graph so that

$$\left\{ \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} = \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}.$$

Artificially construct a coupling of random walk X and GFFs η and η' on the metric graph so that

$$\left\{ \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2 \right\}_{x \in V} = \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}.$$

- Let

$$U = \{\text{set on which } \mathcal{L}_{\tau+(t)}(x) > 0\}.$$

Claim: U is (a.s.) connected.

Artificially construct a coupling of random walk X and GFFs η and η' on the metric graph so that

$$\left\{ \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2 \right\}_{x \in V} = \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}.$$

- Let

$$U = \{\text{set on which } \mathcal{L}_{\tau+(t)}(x) > 0\}.$$

Claim: U is (a.s.) connected.

- $\eta'_x + \sqrt{2t} = 0$ forces $\mathcal{L}_{\tau+(t)}(x) = 0$

Artificially construct a coupling of random walk X and GFFs η and η' on the metric graph so that

$$\left\{ \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2}\eta_x^2 \right\}_{x \in V} = \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}.$$

- Let

$$U = \{\text{set on which } \mathcal{L}_{\tau+(t)}(x) > 0\}.$$

Claim: U is (a.s.) connected.

- $\eta'_x + \sqrt{2t} = 0$ forces $\mathcal{L}_{\tau+(t)}(x) = 0$
- $\eta'_x + \sqrt{2t}$ can't change signs on U and is positive at $x = v_0$

Theorem (generalized Ray-Knight)

$$\left\{ \mathcal{L}_{\tau+(t)}(x) + \frac{1}{2} \eta_x^2 \right\}_{x \in V} \stackrel{law}{=} \left\{ \frac{1}{2} \left(\eta'_x + \sqrt{2t} \right)^2 \right\}_{x \in V}$$

- Only known proofs are by moment calculations. Can we give an explicit coupling?
- Can be understood relatively well when graph is a path or tree. What about a cycle?

- J. Ding. Asymptotics of cover times via Gaussian free fields: Bounded-degree graphs and general trees. *Annals of Probability* **42** (2), 464–496 (2014).
- J. Ding, J. Lee, and Y. Peres. Cover times, blanket times, and majorizing measures. *Annals of Mathematics* **175** (3), 1409–1471 (2012).
- T. Lupu. From loop clusters and random interlacement to the free field. Preprint arXiv:1402.0298.
- M. B. Marcus and J. Rosen. Markov Processes, Gaussian Processes, and Local Times. Cambridge Studies in Advanced Mathematics **100**. Cambridge Univ. Press (2006).
- A. Zhai. Exponential concentration of cover times. Preprint arXiv:1407.7617.