

Critical Branching Random Walks, Branching Capacity and Branching Interlacements

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THU-PKU-BNU Probability Webinar

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- 1 Introduction
- 2 Branching Capacity and Visiting Probability
- 3 Branching Interlacements and Local Limits of Critical Branching Random Walks in Tori
- 4 Other Relations between Branching Capacity and Critical Branching Random Walks
 - The range of branching random walks
 - Cover times of tori by branching random walks
 - Branching recurrence and branching transience
- 5 Open Questions

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Branching Random Walks

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Branching Random Walk { **Branching Process (GW-Process)**
Random Walk

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- At time 0, a particle is located at x .
- Suppose that, at each time n , a particle v is located at $\mathcal{S}_x(v)$.
- At time $n + 1$, v gives birth to a random number, distributed according to μ , of children, and dies afterwards. Each child then moves to a new position $\mathcal{S}_x(v) + Y$, where Y is distributed according to θ . Different particles behave independently.

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In addition, we need some finite moment assumptions on μ, θ .

Background I: Global and Local

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- **Scaling Limit:** Super Brownian Motion, Brownian Snake.
- **Discrete Behavior:** [Today's topic](#)

Background II: Range of CBRW

Le Gall and Lin have established the following result about the range of critical branching random walk conditioned on the total number of offsprings being n , denoted by R_n :

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Le Gall and Lin have established the following result about the range of critical branching random walk conditioned on the total number of offsprings being n , denoted by R_n :

$$\frac{1}{n} R_n \xrightarrow{P} c_1 \quad \text{as } n \rightarrow \infty, \text{ when } d \geq 5;$$

$$\frac{\log n}{n} R_n \xrightarrow{L^2} c_2 \quad \text{as } n \rightarrow \infty, \text{ when } d = 4;$$

$$n^{-d/4} R_n \xrightarrow{d} c_3 \lambda_d(\text{supp}(\mathcal{I})) \quad \text{as } n \rightarrow \infty, \text{ when } d \leq 3;$$

where c_i are some constants and $\lambda_d(\text{supp}(\mathcal{I}))$ stands for the Lebesgue measure of the support of the random measure on \mathbb{R}^d known as Integrated Super-Brownian Excursion.

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In the subcritical dimensions, they also establish the asymptotics of the hitting probability of a distant point by critical branching random walk:

$$\lim_{x \rightarrow \infty} \|x\|^2 \cdot P(\mathcal{S}_x \text{ visits } 0) = \frac{2(4-d)}{d\sigma^2}.$$

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- The asymptotic of $P(\mathcal{S}_x \text{ visits } 0)$ in other dimensions ($d \geq 4$);
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- The range of branching random walk with a general initial configuration.

We answer the first two questions.

We show that: when $d \geq 5$,

$$\lim_{x \rightarrow \infty} \|x\|^{d-2} \cdot P(\mathcal{S}_x \text{ visits } K) = a_d c(K);$$

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and

$$\frac{\log n}{n} R_n \xrightarrow{P} \frac{16\pi^2 \sqrt{\det Q}}{\sigma^2}.$$

Summary

$$P(\mathcal{S}_x \text{ visits } K) \sim \begin{cases} a_d \text{BCap}(K) / \|x\|^{d-2}; & \text{when } d \geq 5; \\ 1/2\sigma^2 \|x\|^2 \log \|x\|; & \text{when } d = 4; \\ 2(4-d)/d\sigma^2 \|x\|^2; & \text{when } d \leq 3. \end{cases}$$

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Note that

$$P(\mathcal{S}_x \text{ visits } K) \sim \begin{cases} a_d \text{Cap}(K) / \|x\|^{d-2}; & \text{when } d \geq 3; \\ 1; & \text{when } d \leq 2. \end{cases}$$

Branching Capacity: Intuition

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$BCap(A) \approx$ How big A is with regard to CBRW

$Cap(A) \approx$ How big A is with regard to RW

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Key Results

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- $\text{BCap}(A)$ is proportional to the asymptotics of the visiting probability of A by CBRW .
- We can construct a natural measure on the set of all 'infinite CBRW trajectories', such that $\text{BCap}(A)$ is the measure of the set of all trajectories that intersects A .

Remark

- *The above results (and many others) are also true for RW and Capacity.*
- *Use the measure above as the intensity measure of PPP, we get the model of 'branching/random interacements'.*

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Capacity for Random Walk

Capacity for Random Walk

For any finite subset K of \mathbb{Z}^d , $d \geq 3$, the escape probability $ES_K(x)$ is defined to be the probability that a random walk starting from $x \in \mathbb{Z}^d$, denoted by $S_x = (S_x(n))_{n \in \mathbb{N}}$, never returns to K .

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$$Cap(K) = \sum_{a \in K} ES_K(a) = \sum_{a \in K} ES_K^-(a).$$

Connection between capacity and visiting probability

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$$\lim_{x \rightarrow \infty} \|x\|^{d-2} \cdot P(S_x \text{ visits } K) = a_d \text{Cap}(K);$$

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Theorem

For any nonempty finite subset K of \mathbb{Z}^d , $d \geq 5$ and $a \in K$, we have

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where τ_K is the first visiting time of K in a Depth-First search and a_d is the same constant as in the random walk case.

Definition of Branching Capacity

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To define the corresponding escape probabilities, we need to introduce adjusted versions of **infinite** branching random walk.

Infinite versions of GW-trees

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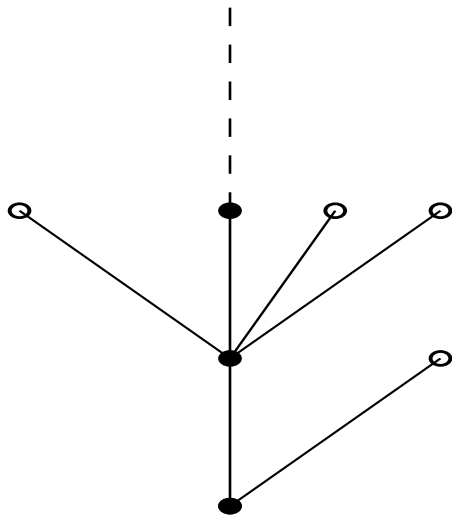
- Each vertex is normal or special; the root is special.
- A normal vertex produces only normal individuals, independently, according to μ .
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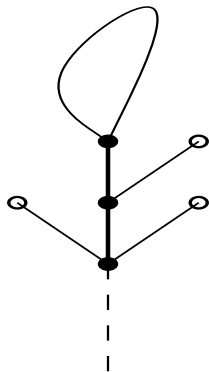
the GW-tree conditioned on survival



What is the local limit around other vertices?

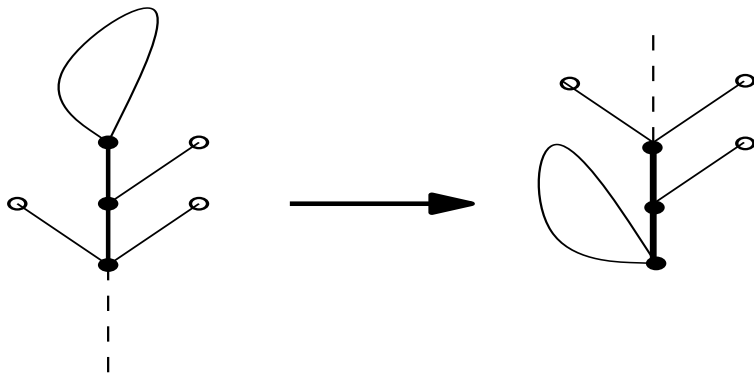
the local limit of GW-trees around other vertices

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- The root produces i individuals with probability $\mu(i - 1)$. The last individual is special; the others (if any) are normal.
- A normal vertex produces only normal individuals, independently, according to μ .
- A special vertex produces individuals independently, according to the size-biased distribution $\tilde{\mu}$, given by $\tilde{\mu}(i) = i\mu(i)$. One of them, chosen uniformly at random, is special; the others (if any) are normal.

Theorem (Aldous)

The local limit of GW-trees (conditioned on the total size) around a uniformly selected vertex is T_v .

T_v and the local limit of GW-trees around any vertex

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The local limit of GW-trees (conditioned on the total size) around a uniformly selected vertex is T_v .

Theorem

The local limit of GW-trees (conditioned on the total size) around any prefixed vertex is T_v , as long as this vertex is not too close to the root.

Invariance Property of T_v

KEY Property

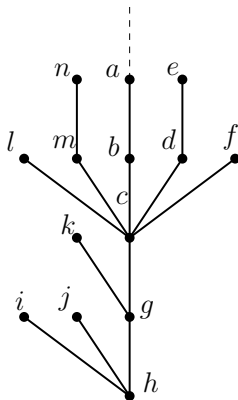
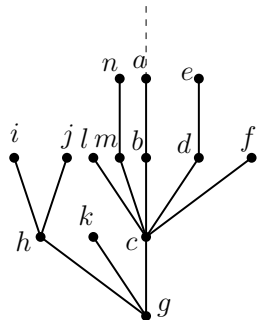
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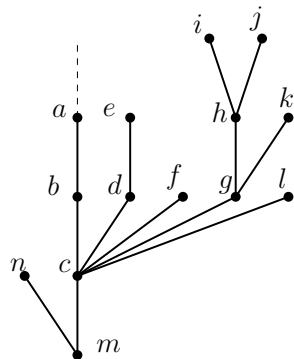
T_v is invariant under τ_v ,

where $\tau_v(t)$ is t re-rooted at the next vertex.

Re-root a tree at different vertices

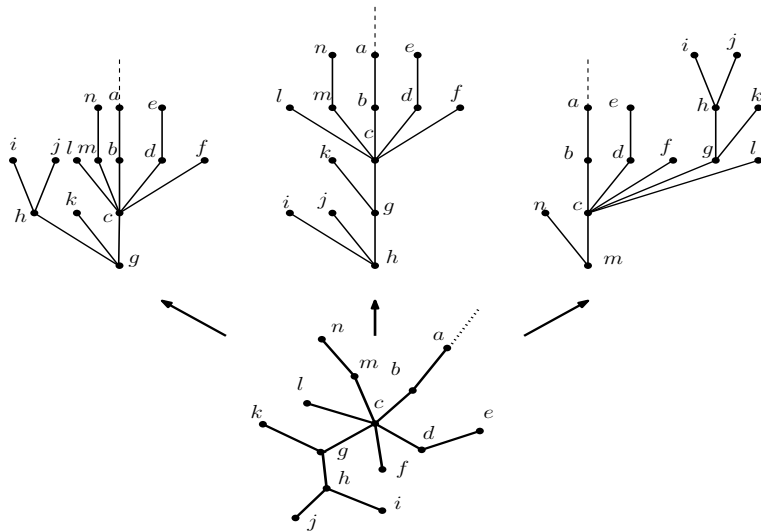


Re-rooted at h



Re-rooted at m

Re-root a tree at different vertices



Definition of escape probability and branching capacity

Now, we can define **branching capacity** for any finite subset of \mathbb{Z}^d ,

Definition

$$Es_A(x) = P[\mathcal{S}_x^v(\text{all vertices before the root}) \cap A = \emptyset];$$

$$Esc_A(x) = P[\mathcal{S}_x^v(\text{all vertices after the root}) \cap A = \emptyset].$$

Then:

$$BCap(A) = \sum_{a \in A} Esc_A(a) = \sum_{a \in A} Es_A(a).$$

Definition of escape probability and branching capacity

Now, we can define **capacity** for any finite subset of \mathbb{Z}^d ,

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$$ES_A^-(x) = P[S_x(\text{all vertices before the root}) \cap A = \emptyset];$$

$$ES_A^+ = P[S_x(\text{all vertices after the root}) \cap A = \emptyset].$$

Then:

$$\text{Cap}(A) = \sum_{a \in A} ES_A^-(a) = \sum_{a \in A} ES_A^+(a).$$

Theorem

For any nonempty finite subset K of \mathbb{Z}^d , $d \geq 5$ and $a \in K$, we have

$$\lim_{x \rightarrow \infty} \|x\|^{d-2} \cdot P(\mathcal{S}_x \text{ visits } K) = a_d \text{BCap}(K);$$

$$\lim_{x \rightarrow \infty} P(\mathcal{S}_x(\tau_K) = a | \mathcal{S}_x \text{ visits } K) = \text{Es}_K(a) / \text{BCap}(K).$$

Moreover, we have

$$\lim_{x \rightarrow \infty} \bar{\Theta}_x \stackrel{d}{=} m_K,$$

where $\bar{\Theta}_x$ is (a random point measure) the conditional entering measure of K by \mathcal{S}_x , conditioned on \mathcal{S}_x visits K .

We also have:

Theorem

For any finite $A \subseteq \mathbb{Z}^d$ and $x \in \mathbb{Z}^d$ with $\rho(x, A) \geq 0.1 \text{diam}(A)$, we have:

$$P(S_x \text{ visits } A) \asymp \frac{BCap(A)}{(\rho(x, A))^{d-2}},$$

where $f(x, A) \asymp g(x, A)$ indicates that there exists positive constants c_1, c_2 independent of x, A such that $c_1 f(x, A) \leq g(x, A) \leq c_2 f(x, A)$.

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Note that for random walk:

$$P(S_x \text{ visits } A) \asymp \frac{Cap(A)}{(\rho(x, A))^{d-2}}.$$

Branching capacity of balls

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Theorem

Let $B^m(r)$ be the m -dimensional balls with radius r (as a subset of \mathbb{Z}^d), i.e. $\{z = (z_1, 0) \in \mathbb{Z}^m \times \mathbb{Z}^{d-m} = \mathbb{Z}^d : |z_1| \leq r\}$, then

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$$BCap(B^m(r)) \asymp \begin{cases} r^{d-4} & \text{if } m \geq d - 3; \\ r^{d-4} / \log r & \text{if } m = d - 4; \\ r^m & \text{if } m \leq d - 5. \end{cases}$$

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One might compare this with the corresponding result about regular capacity:

$$Cap(B^m(r)) \asymp \begin{cases} r^{d-2} & \text{if } m \geq d-1; \\ r^{d-2}/\log r & \text{if } m = d-2; \\ r^m & \text{if } m \leq d-3. \end{cases}$$

$$\text{BCap}(\bullet) = \text{BCap}_{\mu, \theta}(\bullet)$$

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We believe:

$$\text{BCap}_{\mu_1, \theta_1}(A) \asymp \text{BCap}_{\mu_2, \theta_2}(A).$$

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We believe:

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Theorem

Suppose that μ_1, μ_2 are two nondegenerate critical offspring distributions with finite second moment. Then, there is a $C = C(\mu_1, \mu_2) > 0$ such that for all finite $A \subseteq \mathbb{Z}^d$,

$$C^{-1} \cdot \text{BCap}_{\mu_1, \theta}(A) \leq \text{BCap}_{\mu_2, \theta}(A) \leq C \cdot \text{BCap}_{\mu_1, \theta}(A).$$

Proof of $\text{BCap}_{\mu_1}(A) \asymp \text{BCap}_{\mu_2}(A)$

In fact, we will show:

$$P_{\mu_1}(\mathcal{S}_x \text{ visits } A) \asymp P_{\mu_2}(\mathcal{S}_x \text{ visits } A).$$

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Recursive formula for $P(\mathcal{S}_x \text{ visits } A)$:

Let $f(t) = 1 - \sum_{k \geq 0} \mu(k)(1-t)^k$.

$$u_0(x) = 1_A(x), \quad u_n(x) = 1, \forall x \in A; \quad u_{n+1}(x) = f(\mathcal{A}u_n(x)) \forall x \notin A,$$

where $\mathcal{A}u(x) = \sum_{z \in \mathbb{Z}^d} \theta(z)u(x+z)$. Then

$$P(\mathcal{S}_x \text{ visits } A) = \lim_{n \rightarrow \infty} u_n(x).$$

Proof of $\text{BCap}_{\mu_1}(A) \asymp \text{BCap}_{\mu_2}(A)$

Let $f_i(t) = 1 - \sum_{k \geq 0} \mu_i(k)(1-t)^k$ ($i = 1, 2$)

$$u_0^i(x) = 1_A(x); \quad u_{n+1}^i(x) = f_i(\mathcal{A}u_n^i(x)) \forall x \notin A,$$

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There exists $C = C(\mu_1, \mu_2) > 1$ such that, for all $t \in [0, 1]$,

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Therefore, we have $u_\infty^1(x) \leq Cu_\infty^2(x)$, i.e.

$$P_{\mu_1}(\mathcal{S}_x \text{ visits } A) \leq CP_{\mu_2}(\mathcal{S}_x \text{ visits } A).$$

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The model of **random interlacement** consists of a countable collection of **doubly infinite random walk trajectories** on \mathbb{Z}^d , $d \geq 3$. This model is constructed via a Poisson point process with intensity measure $u\nu_0$. The measure ν_0 is supported on the space of **doubly infinite nearest-neighbour trajectories modulo time shifts** on \mathbb{Z}^d . The union of the range of trajectories contained in the support of this Poisson point process defines the random subset I^u of \mathbb{Z}^d , called the **random interlacement** at level u .

The model of **branching interlacement** consists of a countable collection of **infinite tree-indexed random walk trajectories** on \mathbb{Z}^d , $d \geq 5$. This model is constructed via a Poisson point process with intensity measure $u\nu$. The measure ν is supported on the space of **infinite tree-indexed random walk trajectories modulo vertex shifts** on \mathbb{Z}^d . The union of the range of trajectories contained in the support of this Poisson point process defines the random subset \mathcal{I}^u of \mathbb{Z}^d , called the **branching interlacement** at level u .

Characterization of branching interlacements

The law of the **branching interlacement** at level u can be characterized as the unique distribution on $\{0, 1\}^{\mathbb{Z}^d}$ such that

Characterization

$$P[\mathcal{I}^u \cap K = \emptyset] = \exp(-u \cdot \text{BCap}(K)), \text{ for every } K \subset\subset \mathbb{Z}^d, \quad (1)$$

where $\text{BCap}(K)$ is the **branching capacity** of K .

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Note that (1) uniquely determines the law of \mathcal{I}^u by the inclusion-exclusion principle:

$$\begin{aligned} P[\mathcal{I}^u \cap K = A] &= \sum_{B \subseteq A} (-1)^{|A \setminus B|} P[\mathcal{I}^u \cap (K \setminus B) = \emptyset] \\ &= \sum_{B \subseteq A} (-1)^{|A \setminus B|} \exp(-u \text{BCap}(K \setminus B)). \end{aligned}$$

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Let N_K be a Poisson random variable with parameter $u \cdot \text{BCap}(K)$, and $(X^j)_{j \geq 1}$ be i.i.d. **branching random walks** with m_K , the '**branching harmonic measure**' from infinity of K as the initial measure.

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Recall

$$P[\mathcal{S}_x \text{ visits } K] \sim \frac{a_d \text{BCap}(K)}{\|x\|^{d-2}}; \quad \lim_{x \rightarrow \infty} \bar{\Theta}_x \stackrel{d}{=} m_K$$

where $\bar{\Theta}_x$ is the conditional entering measure of K by \mathcal{S}_x , conditioned on \mathcal{S}_x visits K .

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Remark

We can regard branching interlacement as the trajectory of the BRWs from **Infinity**.

Constructive definition of Branching interlacements

$$W = \{(t, \mathcal{S}) : t \in \mathbb{T}_\infty, \mathcal{S} : t \rightarrow \mathbb{Z}^d\};$$

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τ is the transformation on W induced by τ_v .

$$w_1 \sim w_2 \text{ iff } \exists n \in \mathbb{Z}; w_1 = \tau^{(n)}(w_2);$$

$$W^* = W / \sim;$$

$$W_K = \{w \in W : \exists n \in \mathbb{Z}, w(n) \in K\};$$

$$H_K(w) = \inf\{n : w(n) \in K\};$$

$$W_K^0 = \{w \in W_K : H_K(w) = 0\}.$$

Construction of the intensity measure

Recall that we have an invariant measure T_v and the corresponding vertex shift τ_v

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- Each vertex, except the root, is normal or special.
- The root produces i individuals with probability $\mu(i - 1)$. The last individual is special; the others (if any) are normal.
- A normal vertex produces only normal individuals, independently, according to μ .
- A special vertex produces individuals independently, according to the size-biased distribution $\tilde{\mu}$, given by $\tilde{\mu}(i) = i\mu(i)$. One of them, chosen uniformly at random, is special; the others (if any) are normal.

Construction of the intensity measure

For any $K \subset\subset \mathbb{Z}^d$, denote by Q_K , the finite measure on W_K such that

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Proposition

For any measurable set $A \subset W$, we have

$$\sum_{x \in \mathbb{Z}^d} P_x^v[A] = \sum_{x \in \mathbb{Z}^d} P_x^v[\tau(A)].$$

KEY FEATURE

$$Q_K[\pi^{-1}(A) \cap W_K] = Q_{K'}[\pi^{-1}(A) \cap W_K], \quad \forall A \subset W^*.$$

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$$\nu[A] = \lim_{K \uparrow \mathbb{Z}^d} Q_K[\pi^{-1}(A)], \quad \forall A \subset W^*.$$

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Definition

The branching interlacement at level u is defined to be

$$\mathcal{I}^u = \mathcal{I}^u(\omega) = \bigcup_{n \geq 0} \text{Range}(\bar{w}_n), \quad \text{where} \quad \omega = \sum_{n \geq 0} \delta_{\bar{w}_n},$$

is distributed as the PPP with intensity measure $u\nu$.

The local limit of branching random walks in tori

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More precisely,

$$\lim_{N \rightarrow \infty} P[\text{Range}(\mathcal{S}_N^n) \cap K = A] = P[\mathcal{I}^u \cap K = A], \forall A \subset K.$$

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 - The range of branching random walks
 - Cover times of tori by branching random walks
 - Branching recurrence and branching transience
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The range of branching random walks

Recall that we have

$$\frac{R_n}{n} \xrightarrow{P} c_1, \text{ when } d \geq 5,$$

where R_n is the range of tree-indexed random walk conditioned on the size being n .

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Note that, for random walk, we have:

$$\frac{R_n}{n} \xrightarrow{P} \text{Cap}(\{0\}), \text{ when } d \geq 3.$$

Cover times by branching random walks

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Answer

The cover time of \mathbb{T}_N is concentrated at $N^d \log N^d / \text{BCap}(\{0\})$.

Theorem

Let $n = n(N)$ be an integer-valued function of N . For any $\epsilon \in (0, 1)$, we have if $n(N) > (1 + \epsilon)N^d \log N^d / \text{BCap}(\{0\})$, then

$$\lim_{N \rightarrow \infty} P[\text{Range}(\mathcal{S}_N^n) = \mathbb{T}_N] = 1;$$

if $n(N) < (1 - \epsilon)N^d \log N^d / \text{BCap}(\{0\})$, then

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Review of Recurrence and Transience

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A subset $K \subset \mathbb{Z}^d$ is called recurrent if

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Suppose $K \subset \mathbb{Z}^d$, $d \geq 3$ and let $K_n = \{a \in K : 2^n \leq |a| < 2^{n+1}\}$.

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$$K \text{ is recurrent} \Leftrightarrow \sum_{n=1}^{\infty} \frac{\text{Cap}(K_n)}{2^{n(d-2)}} = \infty.$$

Definition

Let K be a subset of \mathbb{Z}^d ($d \geq 5$). We call K a **branching recurrent** (**B-recurrent**) set if

$$P(\mathcal{S}_0^\infty \text{ visits } K \text{ infinitely often}) = 1,$$

and a **branching transient** (**B-transient**) set if

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Therefore, K is B-recurrent.

Branching recurrence of \mathbb{Z}^4

By projection to \mathbb{Z}^4 , we get that 0 is visited infinitely often by the incipient infinite snake, almost surely.

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Corollary

The incipient infinite snake in \mathbb{Z}^4 almost surely visits any vertex infinitely often.

$$\text{BCap}_{\mu_1, \theta}(\bullet) \asymp \text{BCap}_{\mu_2, \theta}(\bullet).$$

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Hence

$$\sum_{n=1}^{\infty} \frac{\text{BCap}_{\mu_1, \theta}(K_n)}{2^{n(d-4)}} = \infty \Leftrightarrow \sum_{n=1}^{\infty} \frac{\text{BCap}_{\mu_2, \theta}(K_n)}{2^{n(d-4)}} = \infty.$$

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Corollary

K is B-recurrent w.r.t. (μ_1, θ) if and only if K is B-recurrent w.r.t. (μ_2, θ) .

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Variational formula for branching capacity and branching harmonic measure

How to characterize $\text{BCap}(A)$ and m_A ?

Note that for random walk capacity we have:

Variational formula for random walk capacity

$$\text{Cap}(A) = \left(\inf \int \int g(x, y) \nu(dx) \nu(dy) \right)^{-1},$$

where \inf is over all probability measures on A , and $g(x, y)$ is the discrete Green function.

Branching Capacity is a $(d - 4)$ -Capacity?

Conjecture

$$\text{BCap}(A) \asymp \left(\inf \int \int (|x - y| \vee 1)^{4-d} \nu(dx) \nu(dy) \right)^{-1},$$

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where inf is over all probability measures on A .

If so, we have

$$\text{BCap}_{\mu, \theta_1}(\bullet) \asymp \text{BCap}_{\mu, \theta_2}(\bullet)$$

Harnack inequality for CBRW

Harnack Inequality for random walk

Suppose $U \subset \mathbb{R}^d$ is open and connected, and K is a compact subset of U . Then there exist $C = C(K, U)$ and $N = N(K, U)$ such that if $n \geq N$,

$$U = U_n = \{x \in \mathbb{Z}^d : n^{-1}x \in U\}, \quad K = K_n = \{x \in \mathbb{Z}^d : n^{-1}x \in K\},$$

we have:

$$h_U(x, a) \leq Ch_U(y, a), \quad x, y \in K, a \in \partial U,$$

where $h_U(x, a) = P(S_x(\tau_{U^c}) = a)$ is the exit measure.

Harnack inequality for CBRW

Harnack Inequality for critical branching random walk (Conjecture)

Suppose $U \subset \mathbb{R}^d$ is open and connected, and K is a compact subset of U . Then there exist $C = C(K, U)$ and $N = N(K, U)$ such that if $n \geq N$,

$$U = U_n = \{x \in \mathbb{Z}^d : n^{-1} \in U\}, \quad K = K_n = \{x \in \mathbb{Z}^d : n^{-1} \in K\},$$

we have:

$$h_U(x, a) \leq Ch_U(y, a), \quad x, y \in K, a \in \mathcal{M}(\partial U),$$

where $h_U(x, a) = P(\Theta_{U^c}(x) = a)$ is the exit measure.

Coupling between Branching interlacements and tree-indexed random walks in tori

For random walks and random interlacements, we have the following coupling result.

Theorem (Teixeria and Windisch)

For any $u > 0, \alpha > 0, \epsilon \in (0, 1)$, there exists a constant c depending on d, u, α, ϵ and a coupling (Ω, \mathcal{A}, Q) of $S = S_{[0, uN^d]}$ with random interlacements $I^{u(1-\epsilon)}$ and $I^{u(1+\epsilon)}$ on \mathbb{Z}^d , such that

$$Q[I^{u(1-\epsilon)} \cap A \subseteq S \cap A \subseteq I^{u(1+\epsilon)} \cap A] \geq 1 - cN^{-\alpha}, \forall N > 1,$$

where A is a box with length $N^{1-\epsilon}$.

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where A is a box with length $N^{1-\epsilon}$.

Question

Can we show some analogous results for branching interlacements?

Cover times

Recall that for $d \geq 5$, the cover time of a d -dimensional torus by critical branching random walk is concentrated at $N^d \log N^d / \text{BCap}(\{0\})$. What about other dimensions?

Recall that for $d \geq 5$, the cover time of a d -dimensional torus by critical branching random walk is concentrated at $N^d \log N^d / \text{BCap}(\{0\})$. What about other dimensions? For the cover time (denoted by C_N) of tori by random walks, we have:

- for $d \geq 3$, $\mathbb{E}[C_N] \sim N^d \log N^d / \text{Cap}(\{0\})$. Moreover $C_N \text{Cap}(\{0\}) / N^d - \log N^d \xrightarrow{d}$ Gumbel distribution (Belius);
- for $d = 2$, $\mathbb{E}[C_N] \sim 4N^2 (\log N)^2 / \pi$ (Dembo-Peres-Rosen-Zeitouni);
- for $d = 1$, C_N / N^2 converges in law to the time needed for BM to cover the 1d-torus with unit length.

Question

Do we have the analogous results for the cover time by CBRW?

The End
Thanks