

Uniform Spanning Trees in High Dimension

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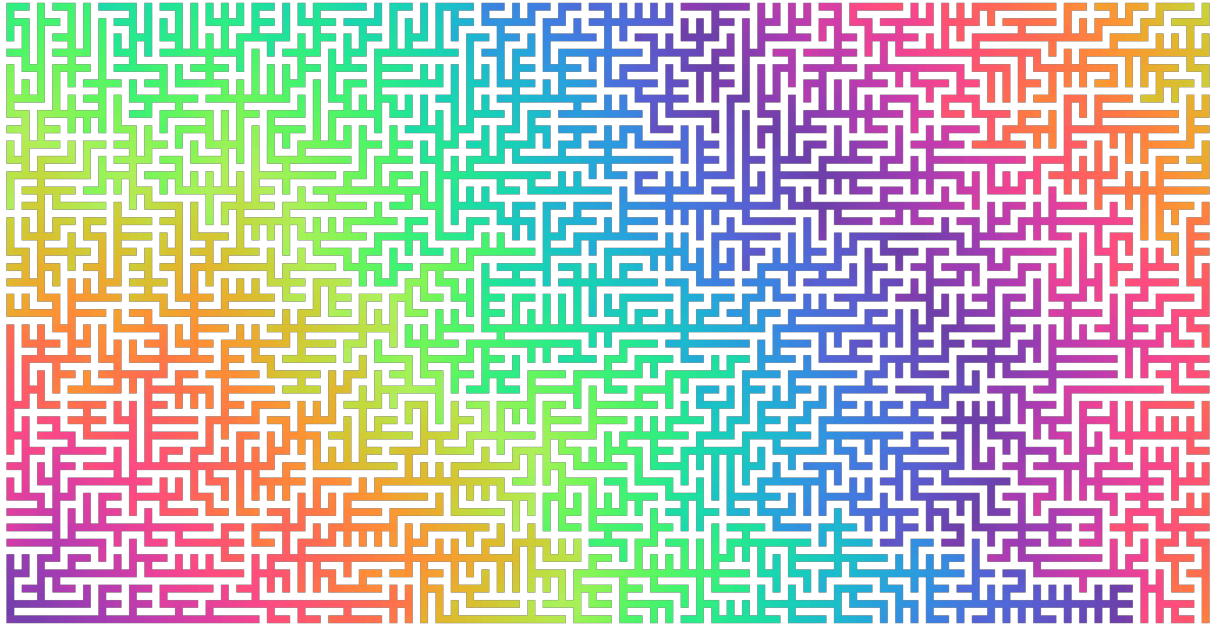


image due to Mike Bostock

Outline of lectures:

- 1: Basics + Sampling Algorithms
- 2: The connectivity / disconnectivity transition
- 3: The interlacement Aldous - Broder Algorithm
- 4: Critical exponents in high dimensions.

§1: Basis + Sampling Algorithms

$G = (V, E)$ finite connected graph

$\mathcal{T}(G)$ set of spanning trees of G .

↑ connected subgraphs of G
containing every vertex and
no cycles.

"Uniform spanning tree" — Uniform random element
of $\mathcal{T}(G)$.

Turns out to be a very interesting object with
connections to many other topics.

△ "statistical mechanics" model that is both
non-trivial and (relatively) tractable

- Development of SLE Schramm,
Lawler, Schramm and
Werner
- 3d scaling limits Kozma,
Angel, Crayden, Hernandez-Torres,
Shiraishi

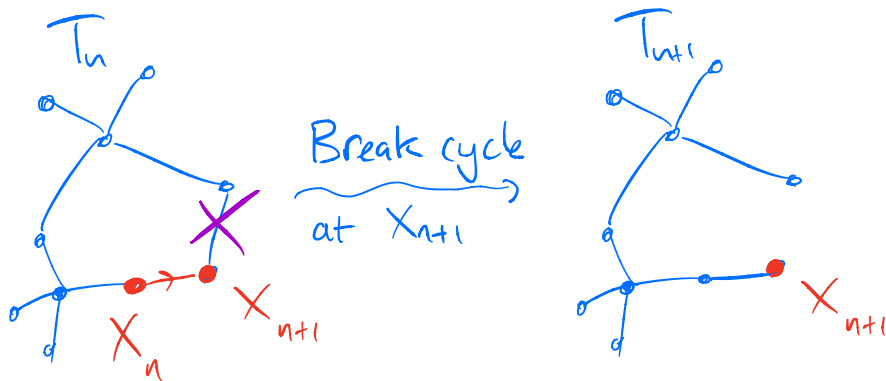
Sampling algorithms

1: MCMC $G = (V, E)$ finite.

Define a Markov chain (X, T) on $V \times \mathcal{T}$ as follows

- X is a random walk
- Given T_n and X_n, X_{n+1} , define T_{n+1} by adding $\{X_n, X_{n+1}\}$ then "breaking the cycle at X_{n+1} "

"UST chain"



(If $\{X_n, X_{n+1}\} \in T_n$, $T_{n+1} = T_n$)

Markov-Chain-Tree Theorem :

$\pi \otimes \text{UST}$ is the unique stationary measure for the UST chain.

stationary measure for SRW

2: Aldous-Broder Algorithm

$G = (V, E)$ finite, ^{connected} $(X_n)_{n \geq 0}$ random walk on G .

$e(v, X)$ = oriented edge crossed by X as it enters v for first time.
not defined for $v \neq X_0$.

$$AB(X) = \{ e(v, X) \leftarrow : v \neq 0 \}$$

↑
reversal

Thm (Broder 1989, Aldous 1990). G finite connected, X random walk started at v . Then $AB(X)$ is distributed as a UST of G oriented towards v .

Proof sketch $(X_n)_{n \in \mathbb{Z}}$ doubly- ∞ SRW, $X_0 \sim \pi$.

$e_n^-(v)$ = last edge used to exit v before time n
 $e_n^+(v)$ = first edge used to enter v after time n
Defined for $v \neq X_n$

Observation:

$(X_n, \{e_n(v) : v \neq X_n\})$ evolves as UST Chain!

Stationarity + MCTT \Rightarrow
 \uparrow by construction.

$(X_n, \{e_n(v) : v \neq X_n\}) \sim \pi \otimes \text{UST}$
 $\left. \vphantom{(X_n, \{e_n(v) : v \neq X_n\})} \right\} \leftarrow$ by reversibility.

$(X_n, \{e_n(v) : v \neq X_n\}) \quad \square$

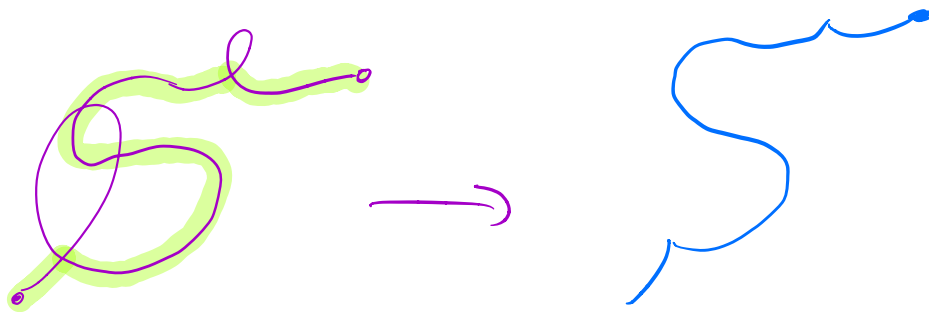
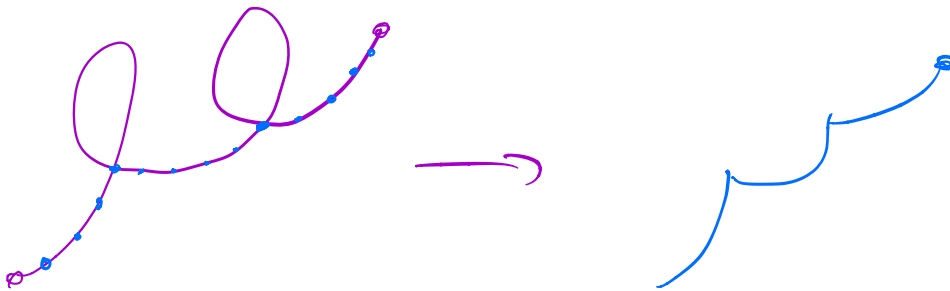
Corollary: Paths in the UST are distributed as loop-erased random walks

LERW: Let $X = (X_n)_{n=0}^N$ either have $N < \infty$ or be transient (visit each vertex at most finitely often).

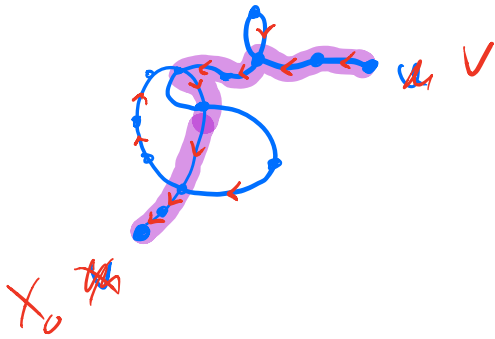
$\ell_0 = 0$ $\ell_{n+1} = 1 + \text{Sup} \{ i \geq \ell_n : X_i = X_{\ell_n} \}$
(Stop when $X_{\ell_n} = X_u$)

$$LE(X) = (X_0, X_{\ell_1}, X_{\ell_2}, \dots)$$

"erase loops chronologically as they are created".



In $AB(X)$, path $X_0 \rightarrow v$
 is equal to $LE((X^{\tau_v})^{\leftarrow})$
 hitting time reversal.



Fact (Lawler): $LE((X^{\tau_v})^{\leftarrow}) \stackrel{d}{=} LE(X^{\tau_v})^{\leftarrow}$

\Rightarrow Corollary.

\uparrow NB: Does not hold pointwise.

3: Wilson's Algorithm

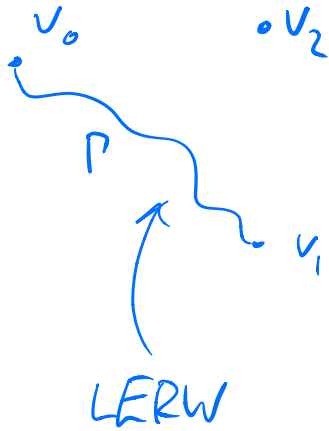
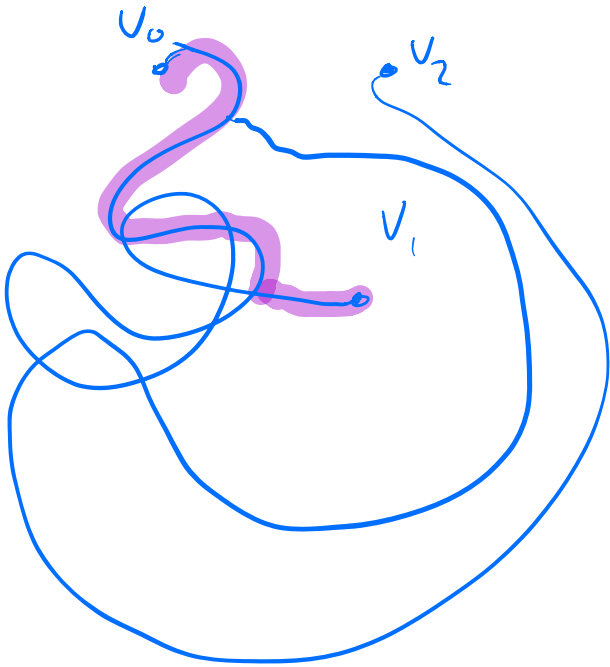
Enumerate $V = \{v_0, v_1, \dots, v_N\}$

$$T_0 = (\{v_0\}, \emptyset)$$

Given T_n , let X^{n+1} be a random walk started at v_{n+1} & stopped when it hits vertex set of T_n .

$T_{n+1} = \text{Union of } T_n \text{ with loop erasure of } X^{n+1}.$

Thm: (Wilson) $T_N \sim \text{UST}$.



Apply Corollary again:

Conditional distr. of path from v_2 to P

$P \in \text{LERW}$

$M \in G/P$

Same thing as LERW in G from v_2 to P !

Markov property of UST:

Conditional distr. of $T \sim \text{UST}$ Given
 $A \subseteq T, T \cap B = \emptyset$

$$\stackrel{d}{=} A \cup \left(\text{UST of } \frac{G-B}{A} \right)$$

delete B , contract A .

There is also a strong spatial Markov property:

Say that a random set $K \subseteq E$
defined on same probability space as T
is a local set if

$\{K \subseteq W\}$ is measurable wrt $\sigma(T|_W)$
 $\forall W \subseteq E$.

E.g. $K =$ path connecting u and v .

Conditional on K and $T|_K$,

$$T \sim K_0 \cup (\text{UST of } \frac{G_0 - K_c}{K_0})$$

↑
open edges
in K
↑
closed
edges
in K .

Wilson's algorithm follows by earlier corollary + induction!

§2: Infinite volume limits and the connectivity/disconnectivity transition.

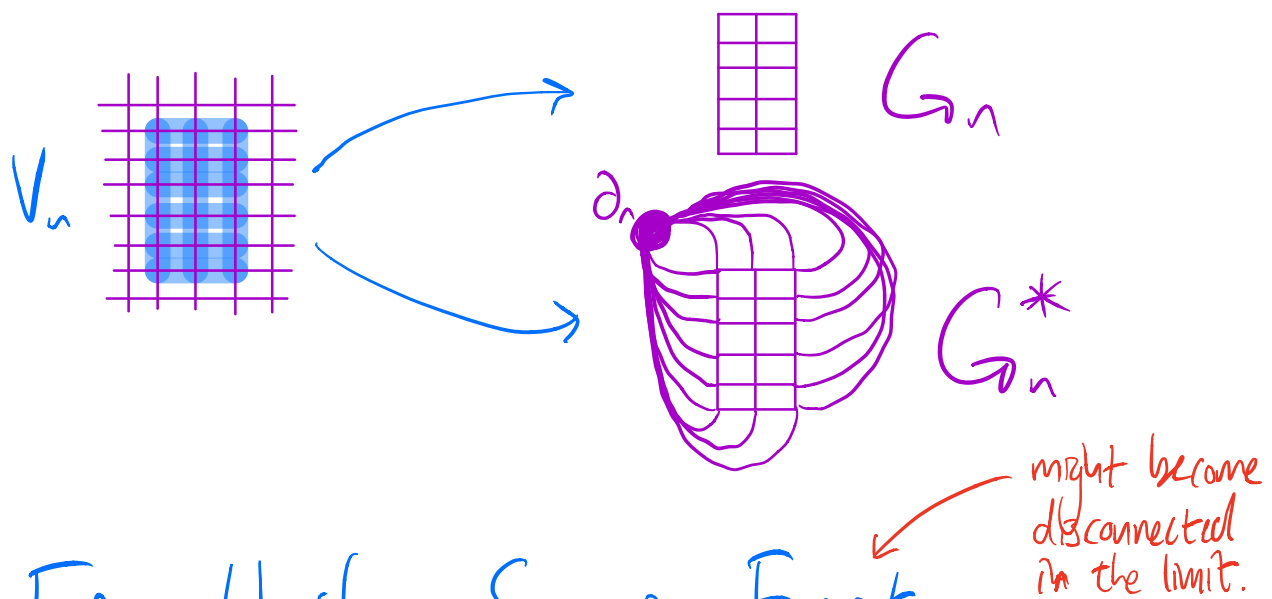
$G_{\infty} = (V, E)$, locally finite, connected graph
 ↑ all degrees finite. (e.g. \mathbb{Z}^d)

$(V_n)_{n \geq 1}$ exhaustion of V by finite connected sets

$$V_n \subseteq V_{n+1} \subseteq \dots, \quad \bigcup_{n \geq 1} V_n = V$$

G_n : subgraph induced by V_n

G_n^* : Contract $V \setminus V_n$ to a single vertex ∂_n



Free Uniform Spanning Forest:

Weak limit of USTs of G_n

Wired Uniform Spanning Forest:

Weak limit of USTs of G_n^*

Both limits exist and are independent of the exhaustion
(Pemantle '91)

Thm: For \mathbb{Z}^d , FUSF = WUSF.

Generally, $FUSF \neq WUSF \Leftrightarrow \exists$ nonconstant harmonic $h: V \rightarrow \mathbb{R}$
 st. $\sum_{x \sim y} (h(x) - h(y))^2 < \infty$

Wilson's algorithm noted at infinity

Benjamini, Lyons, Peres, Schramm

G infinite, transient. $V = \{v_1, v_2, \dots\}$

$\mathcal{F}_0 =$ empty forest with no vertices or edges

In particular, if G is an amenable transitive graph, then $FUSF = WUSF$

G Cayley $\Rightarrow FUSF \neq WUSF \Leftrightarrow \rho$ has positive first L^2 Betti number

$\chi^2(\rho) = \frac{1}{2} (E d_{FUSF}(\rho) - \text{deg}_{\text{normal}}(\rho))$

Given \mathcal{F}_n , let X^{n+1} be a random walk started at v_{n+1} and stopped when it hits vertex set of \mathcal{F}_n , running forever if this never happens.

Let $\mathcal{F}_{n+1} = \mathcal{F}_n \cup LE(X^{n+1})$.

Then $\mathcal{F} = \bigcup_{n \geq 1} \mathcal{F}_n \sim WUSF$

Connectivity / Disconnectivity

Key observation: When running Wilson's algorithm, a new component is formed exactly when X^{n+1} does not hit F^n .

Thm (Pemantle '91) The USF of \mathbb{Z}^d is connected when $d \leq 4$ and has ∞ many components when $d > 4$

Thm (Erdős-Taylor '60s): Two simple random walks on \mathbb{Z}^d intersect infinitely often a.s. when $d \leq 4$, and at most finitely often a.s. when $d > 4$.

High-dimensional part of Erdős-Taylor is easy:

X, Y random walks started at $x, y \in \mathbb{Z}^d$.

$$\mathbb{E} |\{n, m \geq 0 : X_n = Y_m\}|$$

$$= \sum_{n, m \geq 0} \sum_{z \in \mathbb{Z}^d} P_n(x, z) P_m(y, z)$$

$$= \sum_{n, m \geq 0} P_{n+m}(x, y) = \sum_{k \geq 0} (k+1) P_k(x, y)$$

Classical estimate: $P_k(x, y) \approx \frac{1}{k^{d/2}} e^{-\Theta\left(\frac{\|x-y\|^2}{k}\right)}$

$$\rightsquigarrow \mathbb{E} |\{n, m \geq 0 : X_n = Y_m\}| \leq \|x-y\|^{-d+4} \text{ when } d > 4.$$

Also shows that this expectation is infinite when $d \leq 4$.

Low-dimensional part X, Y both start at 0

$$I_N = |\{0 \leq n, m \leq N : X_n = Y_m\}|$$

Calculation we just did shows that

$$E I_N = \sum_{n, m=0}^N p_{n+m}(0,0) \asymp \begin{cases} N^{4-d/2} & d < 4 \\ \log N & d = 4 \\ \mathbb{I} & d > 4 \end{cases}$$

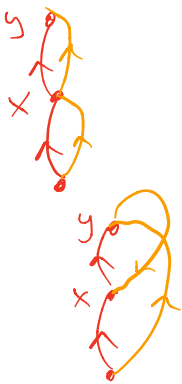
In particular, $E I_N \xrightarrow{N \rightarrow \infty} \infty$ when $d \leq 4$.

On the other hand, letting $\xi_N(x,y) = \sum_0^N p_n(x,y)$

$$E I_N^2 \leq \sum_{x,y} (\xi_N(0,x) \xi_N(x,y) + \xi_N(0,y) \xi_N(y,x))^2$$

$$\leq 2 \sum_{x,y} (\xi_N(0,x)^2 \xi_N(x,y)^2 + \xi_N(0,y)^2 \xi_N(y,x)^2)$$

$$= 4 \sum \xi_N(0,x)^2 \xi_N(x,y)^2 = 4(E I_N)^2$$



Paley-Zygmund:

$$P(I_N \geq \frac{1}{2} \mathbb{E}I_N) \geq \frac{1}{4} \frac{(\mathbb{E}I_N)^2}{\mathbb{E}(I_N^2)} \geq \frac{1}{16}$$

Since $\mathbb{E}I_N \rightarrow \infty$ as $N \rightarrow \infty$, Fatou implies

$$P(I = \infty) \geq \limsup P(I_N \geq \frac{1}{2} \mathbb{E}I_N) \geq 1/16.$$

Hewitt-Savage 0-1 law implies $\{I = \infty\}$ is a 0-1 event \square .

High-d part of Pemantle:

Start k RWs all far away from each other. Then none of these walks intersect each other whp.

\Rightarrow USF has at least k components whp.

Low d part:

transient

Thm (Lyons, Peres, Schramm.) Let X, Y be random walks on some graph. Then

$$\mathbb{P}(Y \cap LE(X) \neq \emptyset) \geq \frac{1}{256} \mathbb{P}(Y \cap X \neq \emptyset)$$

Moreover, if X and Y intersect i.o. a.s. then so do Y and $LE(X)$.

Exercise Let $A \subseteq \mathbb{Z}^3$ be an ∞ connected set. Prove that SRW on \mathbb{Z}^3 intersects A infinitely often almost surely.

§3: One-endedness via interlacement Aldous-Broder.

A tree is said to be one-ended if it does not contain a bi-infinite simple path.

Thm (BLPS) Every tree in the USF of \mathbb{Z}^d is one-ended a.s. for every $d \geq 2$.

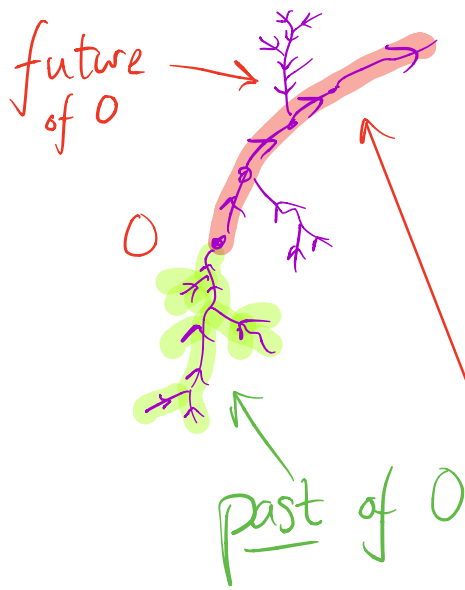
We'll prove the $d=3$ case.

There are now many proofs available, applying at different levels of generality.

I will show an unpublished proof that leads naturally to the critical exponent calculations in high- d .

NB: Oriented WUSF: Orient the UST of G_n^* towards ∂_n before taking limit.

\Leftrightarrow Oriented version of Wilson's algorithm



Every vertex u has a unique oriented path emanating from it, called the future of u .

$$v \in \text{future of } u \iff u \in \text{past of } v.$$

Future distributed as LERW by Wilson.

One-endedness \iff Every vertex has finite past.

Critical exponents: E.g.

$$\mathbb{P}(\text{Past} \geq n) \approx n^{-1/8}$$

$$\mathbb{P}(\text{Past reaches distance } r) \approx r^{-1/8}$$

(We'll have much more to say about this later.)

§ 3.1 The Interlacement Aldous-Broder algorithm

Random interacements:

- Introduced by Sznitman
- "Poissonian soup of doubly-infinite random walk trajectories".

Let G be an ∞ , locally finite graph.

For each $-\infty \leq n \leq m \leq +\infty$, let

$W(n, m) =$ set of transient paths in G
↑ indexed by $[n, m]$.
visits each vertex at most finitely often

$W = \bigcup_{-\infty \leq n \leq m \leq \infty} W(n, m)$ set of all transient paths.

The topology on W :

For each $K \subseteq V$ finite, let $W_K \subseteq W$ be set of paths hitting K .

For each $w \in W_K$, let w_K be the portion of w between first and last visits to K .

The topology on W is generated by the subbasis of open sets

$$\left\{ \{ w \in W : w \in W_K, w_K = w'_K \} : \begin{array}{l} K \subseteq V \text{ finite,} \\ w'_K \in W_K \end{array} \right\}$$

(This is ^{stronger than} ~~not~~ the product topology!)

This topology makes W into a Polish space.

- First and last hitting times, local times at vertices, evaluation at a time all continuous.

$W^* = W / \sim$ where $w_1 \sim w_2$ if $\exists k$ st. $w_1(i) = w_2(i+k) \forall i$

W^* is equipped with the quotient topology, which makes it a Polish space.

Elements of W^* are called trajectories

The random interlacement process is a Poisson point process on $W^* \times \mathbb{R}$.

time

The intensity measure:

- For each $w \in W(n, m)$, define $w^{\leftarrow} \in W(-m, -n)$ to be the reversal of w , $w^{\leftarrow}(i) = w(-i)$.

For each $K \subseteq V$ finite, define a measure Q_K on W_K by

$$\begin{aligned}
 & \text{arbitrary Borel subsets of } W \\
 & Q_K(w|_{(-\infty, 0]} \in A, w|_{[0, \infty)} \in B, w(0) = u) \\
 & = \mathbb{1}(u \in K) \deg(u) P_u(X^{\leftarrow} \in A, \tau_K^+ = \infty) P_u(X \in B) \\
 & \quad \uparrow \text{law of SRW} \quad \uparrow \text{first positive hitting time}
 \end{aligned}$$

$\forall u \in V$ and $A, B \subseteq W$ Borel.

$\pi: W \rightarrow W^*$ projection

$W_k^* = \pi(W_k) = \text{trajectories using } k.$

Thm (Sznitman/Texeira) $G \infty$, locally finite, transient. Then there exists a unique locally finite measure Q^* on W^* such that

$$Q^*(A \cap W_k^*) = Q_k(\pi^{-1}(A))$$

for every $A \subseteq W^*$ and $k \subseteq V$ finite.

Q^* : interlacement intensity measure.

Defⁿ: The random interlacement process on G is the Poisson point process on $W^* \times \mathbb{R}$ with intensity

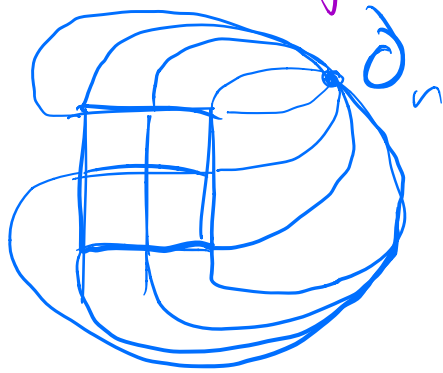
$$Q^* \otimes \text{Lebesgue}.$$

The limiting construction

$G = (V, E)$ ∞ , locally finite, transient.

$V = \cup V_n$ exhaustion by finite sets,

G_n^* defined as before w/ boundary vertex ∂_n



- Take an intensity $\deg(\partial_n)$ Poisson point process on \mathbb{R} .
- For each time t in the process, let W_t be a random walk excursion from ∂_n to itself, which we consider as an element of W^* .

- Get a random set

$$\mathcal{I}_n = \{ (W_t, t) \} \subseteq W^* \times \mathbb{R}$$

Prop \mathcal{I}_n converges weakly to the random interlacement process on G .

In other words:

- Take doubly ∞ SRW on G_n^* .
- Break up into excursions from the boundary.
- Apply Poissonian time change. Scale time in such a way that we get a non-degenerate limit.

Interacements \longleftrightarrow local picture of SRW on the time scale that it covers the graph.

The interlacement Aldus-Broder algorithm

$G = (V, E) \infty$, locally finite, ~~transitive~~
graph. transient

\mathcal{I} random interlacement process on G .

For each $v \in V$ and $t \in \mathbb{R}$,

$\sigma_t(v)$ = First time after time t that v
is hit by a trajectory $W_{\sigma_t(v)}$

$e_t(v)$ = Oriented edge traversed by $W_{\sigma_t(v)}$
as it enters v for the first time.

Thm (H. 2015)

$$AB_t(\mathcal{I}) = \{ e_t(v) \leftarrow : v \in V \}$$

is distributed as the oriented wired uniform
spanning forest for each $t \in \mathbb{R}$

$AB_t(\mathcal{I}_n) \sim \text{UST of } G_n^*$
 by usual Aldous-Broder
 $\implies \text{STP } AB_t(\mathcal{I}_n) \xrightarrow[n \rightarrow \infty]{\text{weakly}} AB_t(\mathcal{I}).$

Although $AB_t: W^* \times \mathbb{R} \rightarrow \{0,1\}^E$ is
 not continuous, it has enough continuity properties
 to prove this by standard arguments (portmanteau
 theorem).

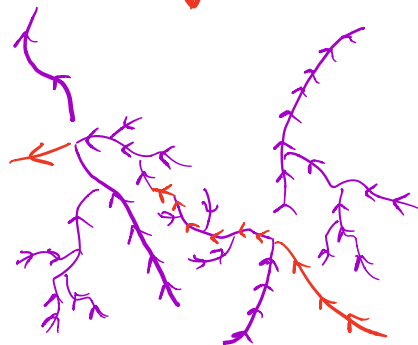
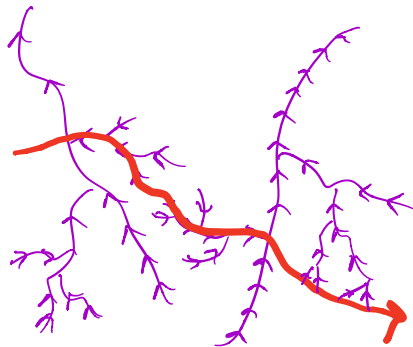
$(AB_t(\mathcal{I}))_{t \in \mathbb{R}}$ is a stationary
 ergodic, stochastically continuous Markov process
 with stationary measure owUST .

This dynamical viewpoint will be of central
 importance to our analysis of the WUST in
 high dimensions.

The past of a vertex evolves in a nice way
under the dynamics:

$$\tilde{\mathcal{F}}_t = AB_t(\mathcal{I}) \quad P_t(v) = \text{past of } v \text{ in } \tilde{\mathcal{F}}_t.$$

As t decreases, new trajectories arrive and overwrite what was there previously



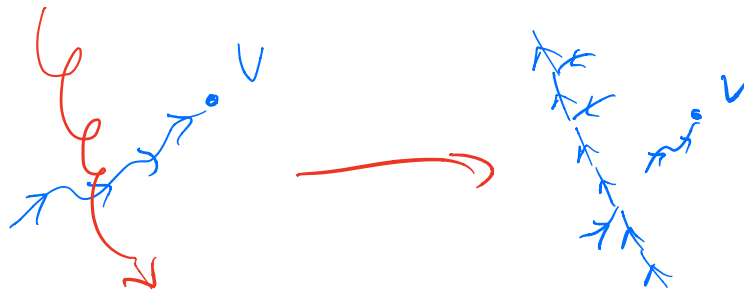
$\tilde{I}_{[s,t)}$ vertices hit in $[s,t)$

Lemma: If $v \notin \tilde{I}_{[s,t)}$ then

$P_s(v) =$ component of v in $P_t(v) \setminus \tilde{I}_{[s,t)}$.

- Let $u \in V$ and let $F_t(u) = (F_t(u)_0 = u, F_t(u)_1, \dots)$ be the future of u in \mathcal{F}_t
- $\sigma_t(F_t(u)_i)$ is decreasing in i .
- $F_s(u)$ and $F_t(u)$ agree until they reach a vertex with $\sigma_s(v) < t$.

After this step, every element of $F_s(u)$ has $\sigma_s < t$.



Interlacement hitting probabilities & the capacity.

Let $K \subseteq V$ be finite. The set of times in which K is visited by an interlacement trajectory is a Poisson process with intensity

$$\begin{aligned} Q^*(W_K^*) &= Q_K(W) \\ &= \sum_{u \in K} \deg(u) P_u(T_K^+ = \infty) \end{aligned}$$

This quantity is known as the capacity or conductance to infinity of K

$$Q^*(W_K^*) = \text{Cap}(K) = \mathcal{E}_{\text{eff}}(K \rightarrow \infty)$$

The theory of electrical networks gives many ways to compute / estimate this quantity.

E.g. "Newtonian capacity" formulation. (Jam & Orey '73)

$$\text{Cap}(K)^{-1} = \inf \left\{ \sum_{u,v \in K} \frac{\xi(u,v)}{\deg(v)} \mu(u)\mu(v) : \mu \text{ a prob. meas. on } K \right\}$$

The Green's function $\xi(u,v) = \sum_{n \geq 0} p_n(u,v)$

For \mathbb{Z}^d , $\xi(x,y) \approx \|x-y\|^{-d+2}$

$$\text{Cap}(K)^{-1} \approx \frac{1}{|K|^2} \sum_{x,y \in K} \|x-y\|^{-d+2}$$

Ball is worst case

$$\text{Cap}(K) \geq |K|^{d-2/d}$$

(cf. isoperimetric inequality $|\partial K| \geq |K|^{d-1/d}$.)

Quantitative one-endedness on \mathbb{Z}^d , $d \geq 3$.

$P_T(v, n) =$ vertices in the past of v with
intrinsic distance at most n
 \uparrow the graph metric on \tilde{T}_T

$$\partial P_T(v, n) = P_T(v, n) - P_T(v, n-1).$$

Theorem Let $d \geq 3$ and consider the USF
of \mathbb{Z}^d . Then

$$P(\partial P_0(0, n) \neq \emptyset) \leq C \frac{\log n}{n^{(d-2)/d}}.$$

Corollary: Every tree is one-ended a.s.

Similar theorem due to Lyons, Morris & Schramm
'03.

Suboptimal result with simple proof.

The mass-transport principle:

If $F: \mathbb{Z}^d \times \mathbb{Z}^d \rightarrow [0, \infty)$ satisfies

$F(x, y) = F(x+z, y+z) \forall x, y, z$ then

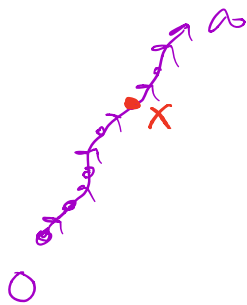
$$\sum_x F(0, x) = \sum_x F(x, 0)$$

E.g. $\mathbb{E} |\partial P_0(0, n)| = 1$ for all n :

$$F(x, y) = \mathbb{P}(y \in \partial P_0(x, n))$$

$$\sum_x F(0, x) = \mathbb{E} |\partial P_0(0, n)|$$

$$\sum_x F(x, 0) = \mathbb{E} |\{x: 0 \in \partial P_0(x, n)\}| = 1.$$



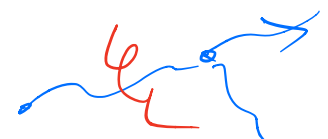
Only one such x , n steps
in the future of 0 .

Proof of Theorem

$$\begin{aligned}
 & \mathbb{P}(\partial P_0(o, n) \neq \emptyset) \overset{O(\varepsilon)}{\leq} \mathbb{P}(\sigma_0(o) \leq \varepsilon) + \mathbb{P}(\sigma_0(o) > \varepsilon, \partial P_0(o, n) \neq \emptyset)
 \end{aligned}$$

Second term

$$\leq \mathbb{P} \left(\begin{array}{l} \partial P_\varepsilon(o, n) \neq \emptyset \text{ and } \exists \text{ a geodesic} \\ 0 \xrightarrow{\varepsilon} \partial P_\varepsilon(o, n) \text{ that is not in } \tilde{\Gamma}_{[0, \varepsilon]} \end{array} \right)$$




(By lemma)

$$= \mathbb{P} \left(\begin{array}{l} \partial P_0(o, n) \neq \emptyset \text{ and } \exists \text{ a geodesic} \\ 0 \rightarrow \partial P_0(o, n) \text{ that is not in } \tilde{\Gamma}_{[-\varepsilon, 0]} \end{array} \right)$$

$$\leq \mathbb{E} \# \left\{ u : u \in \partial P_0(o, n), \begin{array}{l} \text{path from } u \text{ to} \\ 0 \text{ not in } \tilde{\Gamma}_{[-\varepsilon, 0]} \end{array} \right\}$$

$$\leq \mathbb{E} \sum_{u \in \partial P_0(o, n)} e^{-\varepsilon \text{Cap}(o \rightsquigarrow u)}$$



$$\leq e^{-c\epsilon n^{(d-2)/d}} \mathbb{E} \#\{u : u \in \partial P_0(0, n)\} \quad \leftarrow = 1$$

$\uparrow \mathcal{F}_0$ and $\mathcal{I}_{[-\epsilon, 0]}$ independent, hitting controlled by capacity.

$$\text{So } \mathbb{P}(\partial P_0(0, n) \neq \emptyset) \leq C\epsilon + e^{-c\epsilon n^{d/2}}$$

$$\text{Take } \epsilon = C' n^{-d/2} \log n \quad C' \text{ large} \Rightarrow \text{claim } \square$$

This proof can be made to work for e.g. all transient transitive graphs, recovering a result of Lyons, Morris, and Schramm.

To improve this inequality we will need a better understanding of the Capacity of LERW

$$\mathbb{P}(\text{Past} \cap \mathbb{Z} \setminus [-r, r]^d) \asymp \frac{1}{r^2} \neq \emptyset$$

§4: Critical exponents for $d \geq 4$.

4.1: The big picture.

General picture for critical stat mech models:

- Tails of interesting random variables governed by power laws. Exponent should depend on dimension but not choice of lattice

E.g. critical percolation

$$P_{pc}(|\text{Cluster of origin}| \geq n) \approx n^{-1/\delta} + o(1)$$

- There is an "upper critical" dimension above which the exponents stabilize at their "mean-field" values.

"Mean-field" \longleftrightarrow same behaviour as on complete graph / binary tree.

\nearrow terminology comes from field theories like Ising model.

Mean-field behaviour means that having a local interaction

$e^{-\beta \sum_{x \sim y} \sigma_x \sigma_y}$ behaves similarly to $e^{-\beta \sum_x \sigma_x \frac{1}{|x|} \sum_y \sigma_y}$

For percolation, $d_c = 6$.

Conjecturally:

$$P_{p_c}(|K_o| \geq n) \approx \begin{cases} n^{-5/4} & d=2 \\ n^{-1/5(d)} & 3 \leq d \leq 5 \\ n^{-1/2} \log n & d=6 \\ n^{-1/2} & d > 6 \end{cases}$$

$d=2$ understood for site perc on triangular lattice
(Smirnov, Lawler, Schramm, Werner, Kesten)

$d > 6$ Hara - Slade '90s

$3 \leq d \leq 6$ totally open.

For the UST, $d_c = 4$.

Theorem (H. 2018) If $d \geq 5$

$$\mathbb{P}(\partial P_0(o, n) \neq \emptyset) \asymp 1/n.$$

Theorem (H. & Sausi 2020+) When $d=4$,

$$\mathbb{P}(\partial P_0(o, n) \neq \emptyset) \asymp \frac{(\log n)^{1/3}}{n}.$$

- Can derive other related exponents once these are known.
- $d=4$ relies on analysis of 4d LERW due to Lawler.
- Related results on scaling limit of DST of torus due to

Peres & Revelle 2004 $d \geq 5$

Schweinsberg 2009 $d=4$

In both cases there is convergence to the CRT, but the scaling is different in 4d.

- Ideas from our high-d proof have also been used to analyze the diameter of the OST on finite high-d graphs by Michaeli, Nachmias & Shalev (2019).

Lower bounds in high dimensions

Take $d \geq 5$ and let \mathcal{I} be the interlacement process on \mathbb{Z}^d .

$$\text{Thm: } \mathbb{P}(\partial P_0(0, n) \neq \emptyset) \geq \frac{c}{n}.$$

- 0 is hit by a unique trajectory in $[0, \frac{c}{n}]$ with probability $\asymp \frac{1}{n}$.
- With constant probability, the parts of the trajectory before and after hitting 0 are disjoint!

By Erdős-Taylor. Needs $d \geq 5$!



- If this were the only trajectory, the past of O would be infinite.
 - If we apply Aldous-Broder to W we get a tree in which the past of O is infinite.

Let η be an infinite simple path in the past of O in $AB(W)$

↑ One can prove that η is unique and is distributed as a conditioned LERW, but we won't need this for now.

- The forest \mathcal{F}_0 includes the part of η up until the first vertex that is visited by a trajectory that arrived before W .
- By the splitting property of Poisson processes, the first n steps of η

are not hit by any earlier trajectory
with prob. at least

$$e^{-\frac{c}{n} \text{Cap}(\text{First } n \text{ steps of } \eta)} \geq c' > 0.$$

$$\text{Cap}(A) = \sum_{a \in A} \deg(a) P_a(T_A^+ = \infty) \approx |A|$$

So we've shown that in high dimensions

$$P(\partial P_0(0, n) \neq \emptyset \mid 0 \text{ hit in } [0, \frac{1}{n}]) \geq c > 0.$$

and hence

$$P(\partial P_0(0, n) \neq \emptyset) \approx \frac{1}{n} \text{ as claimed. } \square$$

Same strategy gives correct answer in \mathbb{Z}^3 & \mathbb{Z}^4 , but is more difficult to implement rigorously.

Two competing effects

- Capacity of n step LERW grows sublinearly
- Probability two random walks don't intersect decays.

In 4d:

Capacity of n step LERW $\asymp \frac{n}{(\log n)^{2/3}}$

Lawler:

$P(\text{SRW avoids an } n\text{-step LERW}) \asymp \frac{1}{(\log n)^{1/3}}$

\rightsquigarrow Take $\varepsilon = \frac{(\log n)^{2/3}}{n}$, get

$$\begin{aligned} P(\partial P_0(0, n) \neq \emptyset) &\geq \frac{(\log n)^{2/3}}{n} \frac{1}{(\log n)^{1/3}} \\ &= \frac{(\log n)^{1/3}}{n} \end{aligned}$$

To do this properly, one needs to know not just the 'typical' order of the capacity, but to have good enough concentration that the large capacity "bad event" has negligible probability.

Upper bounds when $d \geq 5$.

Want to prove $Q(n) := \mathbb{P}(\partial P_o(o, n) \neq \emptyset) \leq \frac{C}{n}$.

Suffices to prove inductive inequality

$$Q(2n) \leq \frac{C}{n} + \frac{1}{4} Q(n)$$

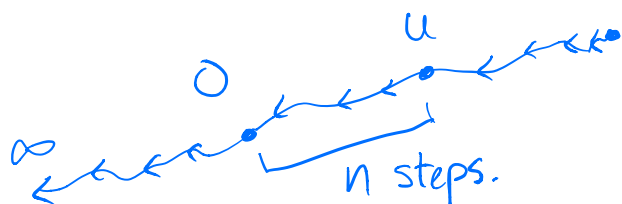
$\swarrow < 1/2$.

Similarly to last time

$$Q(2n) \leq \mathbb{P}(\sigma_o(o) \leq \varepsilon) +$$

$\swarrow = O(\varepsilon)$

$$\mathbb{E} \# \left\{ u : \begin{array}{l} u \in \partial P_\varepsilon(o, n), \text{ path from } \\ u \text{ to } v \text{ in } \mathcal{F}_\varepsilon \text{ not hit in } [0, \varepsilon), \\ \text{AND } \partial P_\varepsilon(u, n) \neq \emptyset \end{array} \right\}$$



Bound by "High Capacity" and "Low Capacity" terms.

$$\mathbb{E} \# \left\{ u : \begin{array}{l} u \in \partial P_\varepsilon(0, n), \text{ path from } \\ u \text{ to } v \text{ in } \hat{f}_\varepsilon \text{ not hit in } [0, \varepsilon), \\ \text{AND } \partial P_\varepsilon(u, n) \neq \emptyset \end{array} \right\}$$

$$\leq e^{-\varepsilon \delta n} \mathbb{E} \# \left\{ u : u \in \partial P_\varepsilon(0, n) \text{ and } \partial P_\varepsilon(u, n) \neq \emptyset \right\}$$

= $Q(n)$ by mass transport!

$$+ \mathbb{E} \# \left\{ u : u \in \partial P_\varepsilon(0, n), \partial P_\varepsilon(u, n) \neq \emptyset, \text{ and } \underline{\text{Cap}}(\overset{\circ}{\circ} \rightsquigarrow u) \leq \delta n \right\}$$

$$\stackrel{(*)}{=} \mathbb{P} \left(\partial P_0(0, n) \neq \emptyset \text{ and first } n \text{ steps of future have } \text{Cap} \leq \delta n \right)$$

Suppose we can show that if δ is sufficiently small then

$$(*) \leq \frac{1}{8} Q(n).$$

Then we get

$$Q(2n) \leq C\varepsilon + e^{-\varepsilon\delta n} Q(n) + \frac{1}{8} Q(n)$$

→ Take δ small, $\varepsilon = \frac{-\log 8}{\delta n}$

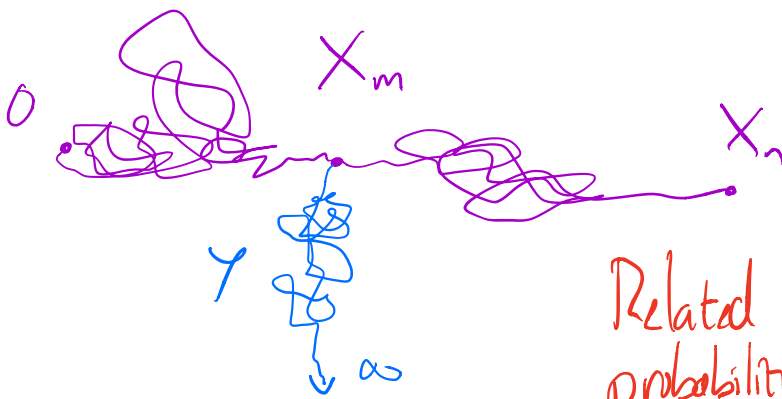
→ Done.

The capacity of SRW & LERW

$$\text{Cap}(A) = \sum \deg(v) P_v(T_A^+ = \infty)$$

X^n first n steps of random walk on \mathbb{Z}^d .

How does $\text{Cap}(X^n)$ grow?



Related to non-intersection probabilities.

$$\mathbb{E} \text{Cap}(X^n) =$$

$$2d \sum_{m=0}^n P\left(Y \text{ does not hit } (X^m \cup \mathbb{Z}^{n-m}) \text{ after time } 0\right)$$

Erdős-Taylor \Rightarrow

$$\mathbb{E} \text{Cap}(X^n) \asymp n \quad \text{when } d \geq 5.$$

Lawler:

$$\mathbb{E} \text{Cap}(X^n) \asymp \frac{n}{\log n} \quad \text{when } d=4$$

Asselah, Shapira, Sousi : Good concentration in both cases.

Easy argument giving lower bound of correct order: $\cup \cup$

$$\text{Cap}(A)^{-1} = \inf \left\{ \sum_{u,v \in A} \frac{\xi(u,v)}{\deg(v)} \mu(u)\mu(v) : \mu \text{ prob meas on } A \right\}$$

$$\asymp \frac{1}{|A|^2} \sum_{u,v \in A} \xi(u,v)$$

\swarrow c.n whp when $d \geq 3$.

$$\longrightarrow \text{Cap}(X^n) \geq \frac{|\{X_m : 0 \leq m \leq n\}|^2}{\sum_{i,j=0}^n \xi(X_i, X_j)}$$

One can compute that

$$\mathbb{E} \sum_{i,j=0}^n g(X_i, X_j)$$

$$\leq 2(n+1) \sum_{m=0}^n (m+1) p_m(0,0) + 2(n+1)^2 \sum_{m=n+1}^{\infty} p_m(0,0)$$

Similar to proof
of Erdős-Taylor

$$\leq \begin{cases} n & d \geq 5 \\ n \log n & d = 4 \\ n^{3/2} & d = 3 \end{cases}$$

Gives lower bound of the correct order for the capacity in each case.

What about LERW?

Earlier this theorem was mentioned:

transient

Thm (Lyons, Peres, Schramm) Let X, Y be random walks on same graph. Then

$$P(Y \cap LE(X) \neq \emptyset) \geq \frac{1}{256} P(Y \cap X \neq \emptyset)$$

Moreover, if X and Y intersect i.o. a.s. then so do Y and $LE(X)$.

With Perla, we show that essentially the same proof gives the following:

Thm (H&Sousi) Let G be a transient graph, and let X be a random walk on G . Then

$$\mathbb{E} \text{Cap}(LE(X^n)) \geq \frac{1}{256} \mathbb{E} \text{Cap}(X^n) \quad \forall n \geq 1$$

In 4d, Lawler shows that

$$|LE(X^n)| \asymp \frac{n}{(\log n)^{1/3}} \text{ whp,}$$

So $\text{Cap}(\text{First } n \text{ steps of LERW})$

$$\approx \text{Cap}(LE(X^{n(\log n)^{1/3}})) \asymp \frac{(\log n)^{2/3}}{n}.$$

This statement is only in expectation. We show that it holds with high probability also. (There is concentration)

In high dimensions it is a much easier matter to get that the capacity of the LERW is linear with high probability.

There is still a problem:

$$(*) \mathbb{P} \left(\begin{array}{l} \partial P_0(0, n) \neq \emptyset \text{ and first} \\ n \text{ steps of future have } \text{Cap} \leq \delta n \end{array} \right)$$

These events aren't independent!

If they were, we could take δ suff
Small and get

$$(*) \leq \frac{1}{8} Q(n) \text{ as required.}$$

Not a problem if

$$\mathbb{P}(\text{Cap}(n \text{ step LERW}) \leq \delta n) \ll \frac{1}{n}$$

This should work when $d \gg 4$, might not
for $d=5$.

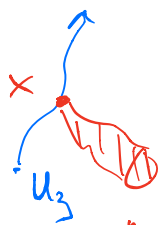
The v -wired uniform spanning forest

Introduced by Jorai & Rediz.

G_n^{*v} : Identify v and ∂_n in G_n^*

v -WUSF is weak limit of USTs of G_n^* oriented towards $\partial_n = v$.

Stochastic domination property (Lyons, Morris, Schramm)



Condition on futures of u_1, \dots, u_k in WUSF or v -WUSF

Conditional distribution of the part of the past of x that does not belong to the revealed forest is stochastically dominated by the component of x in the x -WUSF.

Plays an analogous role to BK inequality in percolation.

In particular, if we define

$$Q_0(n) = \mathbb{P}(\text{o-WUSF has intrinsic radius at least } n)$$

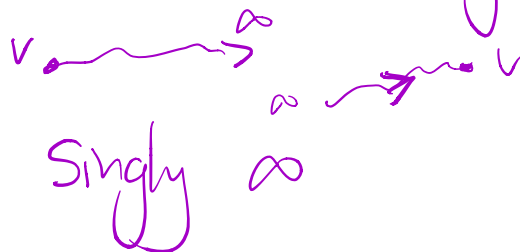
Then $Q(n) \leq Q_0(n) \quad \forall n$.

There are analogues of Wilson & Interlacement Aldous-Broder for the v -WUSF.

Wilson same as before but take

$$\tilde{F}_0 = (\mathbb{Z} \cup 3, \emptyset)$$

" v -wired interlacement" includes trajectories



Otherwise similar to before.

$$Q_0(2n) \leq \mathbb{P}(\sigma_0(0) \leq \varepsilon) \leftarrow \text{still } O(\varepsilon)$$

$$\mathbb{E} \# \left\{ u : \begin{array}{l} u \in \partial P_\varepsilon(0, n), \text{ path from } \\ u \text{ to } v \text{ in } \mathcal{F}_\varepsilon \text{ not hit in } [0, \varepsilon), \\ \text{AND } \partial P_\varepsilon(u, n) \neq \emptyset \end{array} \right\}$$

All forests, interacements are now 0-wired

Split into high capacity and low capacity trees as before.

$$\leq e^{-\varepsilon \delta n} \mathbb{E} \# \left\{ u : u \in \partial P_\varepsilon(0, n) \text{ and } \partial P_\varepsilon(u, n) \neq \emptyset \right\}$$

$$\leq Q_0(n) \mathbb{E} \# \left\{ u : u \in \partial P_\varepsilon(0, n) \right\}$$

by stochastic domination.

$$+ \mathbb{E} \# \left\{ u: u \in \partial P_\varepsilon(0, n), \partial P_\varepsilon(u, n) \neq \emptyset, \right. \\ \left. \text{and } \text{Cap}(\overset{u}{\curvearrowright}) \leq \delta n \right\}$$

$$\leq Q_0(n) \mathbb{E} \# \left\{ u: u \in \partial P_0(0, n) \text{ and } \text{Cap}(\overset{u}{\curvearrowright}) \leq \delta n \right\}$$

New problem: no mass-transport!

Luckily, fairly simple analysis gives

$$\mathbb{E} \# \{ u: u \in \partial P_0(0, n) \} \leq C$$

ln high dim.

$$\mathbb{E} \# \left\{ u: u \in \partial P_0(0, n), \text{Cap} \leq \delta n \right\} \leq \xi_\delta$$

\uparrow
 $\rightarrow 0$
 as $\delta \downarrow 0$.

Problem : For $d \leq 4$,

$$Q_0(n) \gg Q(n).$$

This makes things much harder!

E.g. in 4d $Q(n) \asymp \frac{(\log n)^{1/3}}{n}$

$$Q_0(n) \asymp \frac{(\log n)^{2/3}}{n}$$

In 3d, the powers are different.

In 2d, $Q_0(n)$ is bounded below!!

Connection to sandpile model (Jorai & Redig)

$$\begin{array}{ccccc} \text{Past of } 0 & \leq & \text{Sandpile} & \leq & \text{Component of } 0 \\ \text{in WUSF} & & \text{Avalanche} & & \text{in } 0\text{-WUSF} \\ \uparrow & & & & \uparrow \end{array}$$

In high dim these have the same asymptotics, so we can obtain good understanding of the sandpile.
For $d \leq 4$ there is a gap!

Open problem: Is

$$Q(n) \geq \frac{C}{n} \quad \text{in every dimension?}$$

Would give $\beta \geq 3/2$ where β is the dimension of 3d LERW.

Numerically $\beta \approx 1.62400\dots$ (Wilson)

Rigorously $1 < \beta \leq 5/3$ (Lawler).

It is known that $Q_0(n) \geq \frac{C}{n} \forall n$