

Random Spanning Trees.

(1) Uniform spanning trees:

Consider a finite connected graph D with (V, E) , where E can contain multiple edges.

Def: i) A spanning tree in D is a subset T of E s.t. (V, T) is a tree which doesn't contain a cycle and uses edges only once, connected.

ii) A uniform spanning tree (UST) in D is a random tree T that chooses uniformly among all spanning trees of D .

Rmk: Note that ST must use up the vertices of D .

Def: For $D \subset \mathbb{Z}^d$. finite subset

i) $\partial D = \{x \in \mathbb{Z}^d \mid \chi(x, D) = 1\}$, $\bar{D} = D \cup \partial D$.

ii) $\bar{E} = \bar{E}_{\bar{D}}$ set of edges at least one extremity in D .

iii) Next, we see all points in ∂D as one abstract point x_0 .

Set $\hat{D} = D \cup \{x_0\}$. with set of edges

\hat{E} . left by replacing $\{x, y\}$ with $\{x, x_0\}$.

if $y \in \partial D$. in E .

iv) Denote this bijection map: $E \rightarrow \hat{E}$ by γ .

① Loop-erased RWs:

For $(z_0, \epsilon_1, z_1, \epsilon_2, \dots, z_m, z_m)$, st. $z_0 = z_m$.

ϵ_i connects z_{i-1} and z_i . $\epsilon_i \in \hat{E}$. $z_i \in \hat{\rho}$.

It's called a loop with length m .

\Rightarrow Denote it by (z_0, z_1, \dots, z_m) simply.

Def: (Loop-erasure on path)

For path $z = (z_0, \dots, z_m) \in (\mathbb{Z}^*)^{m+1}$.

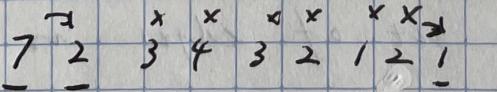
Loop erasure of z is $L(z) = (L_0 \dots L_r)cz$

Defined iteratively: $L_0 = z_0$. $r_j = \min\{i \leq m \mid$

$z_r = L_j\}$. $L_{j+1} = z_{r_j+1} \dots \sigma = \min\{j \mid L_j = z_m\}$

Rmk: i) We erase loop of z chronologically

ii) The edge from L_j to L_{j+1} is the edge from z_{r_j} to $z_{r_{j+1}}$ in z .

Ex. 

Lemma: For path $z = (z_1, z_2, \dots, z_n)$ of SRW.

z is exit time of D . $z \in D$. Then.

$$\Pr(L(z) = (\eta_0, \dots, \eta_s)) =$$

$$(2\lambda)^{-s} h_0(\eta_0, \eta_1) h_0(\eta_1, \eta_2) \dots h_0(\eta_{s-1}, \eta_s)$$

$$\underline{\text{Pf: LHS}} = (2\lambda)^{-s} \prod_{k_i \geq 0} (2\lambda)^{-k_i} \# \text{ paths } \gamma_{i-1} \rightarrow \gamma_i$$

with k_i steps in $D/\{\gamma_0, \dots, \gamma_{i-2}\}$.

$(2\lambda)^{-s}$ is prob. of jump of each step.

$\sum \Delta$: meas the prob. of erased loops from $\gamma_i \rightarrow \gamma_i$ with finite length k_i .

② Wilson's Algorithm:

Next, we construct random tree in $D = [x_i]$.

i) (Z_n) is SRW starting at x_1 . Z is the first exit of D . Let $Z = (Z_k)_{k \leq 2}$.

Consider Loop-erasure $L(Z) \stackrel{a}{=} X^{(1)}$

$$\text{iii) Set } X^{(k)} = \begin{cases} Z & \text{if } Z \in X^{(k-1)} \\ L(\tilde{Z}) \cup X^{(k-1)} & \text{if } Z \notin X^{(k-1)} \end{cases}$$

where \tilde{Z} is SRW starting at x_k and stop

when it first exits $D/X^{(k-1)}$.

iteratively for each $k \in \{2, 3, \dots, n\}$.

iii) We obtain $L(X^{(n)}) = T$ is random tree of \hat{D} .

Pf. Law of T is UST of \hat{D} .

Pf: i) For n outcome T for tree $X^{(n)}$.

First, relabel the points of D :

Denote path of SRW from x_0 to ∂D

by $\gamma_1 \gamma_2 \dots \gamma_{s^1}$.

Next choose the smallest index j . St.

$x_j \in \Sigma \gamma_i \beta_i^{s^1}$. generate a SRW begin
at x_j until hitting $\partial D \cup \Sigma \gamma_i \beta_i^{s^1}$.

Denote the path by $\gamma_s \dots \gamma_{s^1}$.

Iteratively until D is exhausted.

\Rightarrow obtain $\gamma_1 \dots \gamma_n$. $\{\gamma_i\} = \{x_i\}$.

2) By Lemma above.

$$\begin{aligned} |P(\gamma = T)| &= (2d)^{-n} \prod_{j=1}^n h_0 / \gamma_1 \dots \gamma_{j-1} \gamma_j \cdot \gamma_j \\ &= (2d)^{-n} |h_0| \end{aligned}$$

doesn't depend on order of $\{\gamma_i\}$.

Cor. Number of spanning trees in \hat{D}

$$= (2d)^n / |\bar{D}_0|.$$

Rmk: i) $|P(\gamma = T)|$ is constructed without any

$$\text{erasing } |\gamma = T| = (2d)^{-n} / |P(\gamma = T)|$$

$$= |\bar{D}_0| = |h_0|.$$

ii) Denote λ is the loop from x_0 to x_1 that erased when performing the algorithm. which fist. like $z =$

(z_0, \dots, z_r) . $z_0 = x_i$. c is the last time at x_i before hitting x_0 firstly.

prop. $\text{IP}(X = (z_0, \dots, z_r)) = (2d)^{-r} / h_0(x_0, x_i)$

Pf: $LHS = \text{IP}_{x_i}(X_0, X_1, \dots, X_r) = (z_0, z_1, \dots, z_r) \cdot \tilde{H}_{(x_0, x_i)} > T_0 \theta r$

$$\stackrel{\text{def}}{=} \text{IP}_{x_i}(X_0, \dots, X_r) = (z_0, \dots, z_r) \cdot \text{IP}_{x_i}(\tilde{H}_{(x_0, x_i)} > T_0)$$

$$= (2d)^{-r} / h_0(x_0, x_i).$$

$\text{IP}_{x_i}(\tilde{H}_{(x_0, x_i)} > T_0) = 1/h_0(x_0, x_i)$ follows from
the last exit id.

③ Generalisations:

i) With killing:

Consider infinite graph $\hat{D}_0 = \hat{D} \cup \{\delta\}$.

endowed with killing measure $K = (k(x_i))_{x_i \in \hat{E}_0}$

Defn: $c_0 = \begin{cases} k(x_i), & c = s(x_i, \delta) \\ 1/2d, & c \subset \hat{E} \end{cases}$

Thm. $\text{IP}(Y = T = (\eta_1, \dots, \eta_n)) = \prod_{c \in T} c_0 \cdot \text{Aut}(h_0)$.

Pf: $LHS = \prod_{c \in T} c_0 \cdot \prod_{k=1}^n h_0(s(\eta_1, \dots, \eta_{k-1}), t(\eta_k, \eta_k))$

ii) General graph:

$D = (V, E)$ is finite connected graph. (λ_i)

is number of neighbour of x_i in D .

Thm. $P(\mathcal{T} = T) = \prod_{i=1}^n \frac{1}{\pi_i} \cdot \text{det } \Lambda_0$

iii) Weighted Spanning Trees:

Def: A c -weighted spanning tree in \mathbb{D} is a random spanning tree T . s.t.

$P(\mathcal{T} = T) = \frac{\prod_{e \in T} c_e}{\sum_T \prod_{e \in T} c_e}$, where $c_e = c(e)$ is transition prob. along e .

Thm Consider $c = c(e)$ is conductance func. on \mathbb{E} .

Then the random tree T constructed by Wilson Algorithm. is a c -weighted spanning tree.

Besides, $P(\mathcal{T} = T) = \prod_{e \in T} c_e \cdot \text{det } \Lambda_0$

Pf: Note $P(\mathcal{T} = T) \propto \prod_{e \in T} c_e$.

Rmk: It's not MST.

④ Conti. Case Wilson Algorithm:

Consider replace SRW in Wilson Algorithm

nbrr by conti. time RW (X_t). With the jump rate $1/c_e$ along each edge and holding time $\sim \text{Exp}(1)$

(2) Occupation Field:

Def: i) Consider Wilson's Algorithm constructing MST of \hat{D} in discrete time. For $x \in D$. Set $V(x)$ is cumulative time spent at x by all SRW in the algorithm.

ii) Similar as above : set $W(x)$ is occupation time spent at x in conti. case.

iii) $V = (V(x))_{x \in D}$, $W = (W(x))_{x \in D}$ are called occupation fields. in Wilson's algorithm.

Rmk: For $(\zeta_{i,j})_{j \in \mathbb{N}, i \geq 1}$ $\stackrel{i.i.d}{\sim} \text{exp}(1)$. indep of V . we have:

$$(\tilde{W}(x_j))_{j=1}^n =: (\sum_{i=1}^{V(x_j)} \zeta_{i,j})_{j=1}^n \stackrel{i.i.d}{\sim} W = (W_i)_0$$

Prop. i) $V(x_i) \sim \text{Exp}(\lambda_{\text{lock}(x_i, x_i)})$.

ii) $W(x_i) \sim \text{Exp}(\lambda_{\text{lock}(x_i, x_i)})$

Pf: i) $\mathbb{E}_{x_i}^c V(x_i) = \mathbb{E}_{x_i}^c \sum_{k \geq 0} \mathbb{I}_{X_k = x_i} = \lambda_{\text{lock}(x_i, x_i)}$

$$\mathbb{P}_{x_i}^c V(x_i) > m+n \mid V(x_i) > m) =$$

$$\mathbb{P}_{x_i}^c V(x_i) > n \mid V(x_i) > m) =$$

$\mathbb{P}_{x_i}^c (V(x_i) > n) \Rightarrow \text{Memoryless prop.}$

ii) Similar as i)

prop. Laplace Transf.)

For $k : D \rightarrow \mathbb{R}^{2^n}$. Then we have

$$\begin{aligned}\mathbb{E} \left[e^{-\sum_{i=1}^n k(x_i) W(x_i)} \right] & \stackrel{1)}{=} \mathbb{E} \left[e^{-\frac{n}{\pi} (1 + k(x_i))^{-V(x_i)}} \right] \\ & \stackrel{2)}{=} \det(-\bar{\Delta}_D) / \det(-\bar{\Delta}_D + \mathbb{I}_K)\end{aligned}$$

where $\mathbb{I}_K = \text{diag}\{k(x_1), \dots, k(x_n)\}$.

Rmk: Note that W and V are indept of the constructed spanning trees \mathcal{T} . Since the order of (x_i) doesn't influence W, V .

$$\begin{aligned}\text{Pf: 1')} \text{ LHS} &= \mathbb{E} \left[\mathbb{E} \left[e^{-\sum_{i=1}^n \sum_{j=1}^n k(x_j) S_{i,j}} | V \right] \right] \\ &= \mathbb{E} \left[\frac{n}{\pi} \mathbb{E} \left[e^{-k(x_i) S_{i,i}} \right] \right] \\ &= \text{RHS}\end{aligned}$$

2) Work on conti-time case:

Coupling \mathcal{T} with killing version \mathcal{T}'

\mathcal{T}' is a spanning tree constructed from Wilson's Algorithm in conti-time RW with killing measure $k = (k(x_i))_{i \in n}$.

Denote $\widetilde{E}_1 = \{x_i, \partial\}_{1 \leq i \leq n}$.

Note $\mathcal{T}' \trianglelefteq \mathcal{T} \iff \mathcal{T}' \cap \widetilde{E}_1 = \emptyset$.

$$\mathbb{P} \subset J' \cap \widetilde{E}_\delta = X \mid J = T) =$$

$\mathbb{P} \subset \mathcal{V}_{XK}, 1 \leq k \leq n$. RW won't be killed in
the interval $W \subset X_{k+1} \mid J = T)$

$$= \frac{n}{n} \mathbb{P} \subset N \text{ kill in } W \subset X_{k+1} \mid J = T)$$

$$= \frac{n}{n} e^{-k \alpha(x_i) W_{X_{k+1}}}$$

$$\Rightarrow \mathbb{P} \subset J' \cap \widetilde{E}_\delta = X) = \mathbb{E} \subset e^{-\sum_i^n k \alpha(x_i) W_{X_{k+1}}}$$

$$Lns = \sum_{T \in \tilde{\Omega}} \mathbb{P} \subset J' \cap \widetilde{E}_\delta = X, J' = T)$$

$$= \sum_{T \in \tilde{\Omega}} \mathbb{P} \subset J' = T)$$

$$= \sum_{\tilde{\Omega}} (2\pi)^{-n} \cdot |\Lambda_{0,k}| = |\Lambda_{0,k}| / |\Lambda_0|$$

Cor. Suppose $(\varphi'), (\varphi'')$ are indept hFF
with Dirichlet boundary in D

$$\Rightarrow (\varphi')^2/2 + (\varphi'')^2/2 \sim w$$

Pf: Directly by Laplace Transf.

Rmk: If we $\widetilde{V}(x_i) = V(x_i) - 1 \geq 0$.

$$\mathbb{P}_{\text{fin}} := \mathbb{E} \subset \frac{n}{n} (1 + k \alpha(x_i))^{-\widetilde{V}(x_i)}, =$$

$$\frac{\frac{n}{n} (1 + k \alpha(x_i))}{1 - \bar{\Delta}_0 + \bar{\Delta}_{0,k}}$$

$$= |\bar{\Delta}_0| / |\bar{\Delta}_{0,k}|.$$