

Announcements

- start recording
- scribe: work w/other scribe to produce one set of notes
- My OH: Mon 12:30-1:30, Thu 9-10
- TF section/OH: Mon 2-3, Tue 8-9, Wed 5-6, Thu 5:30-6:30
- PS 4 posted. Project proposal due Sun 11/8, problems due TUE 11/17
- If Zoom gets down, check Piazza
- sync whiteboard

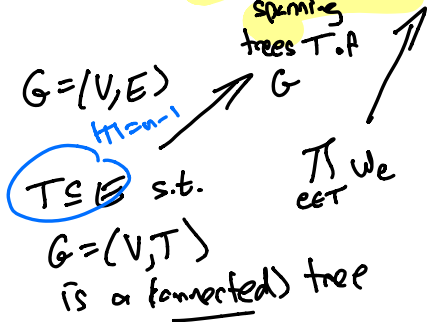
Agenda

- Matrix-Tree Theorem
- Leverage Scores

Matrix - Tree Theorem

Let G be a connected, weighted, undirected graph with Laplacian L .

Then $n \cdot \sum_{\text{spanning trees } T \in \mathcal{T}} w(T) = \sigma_{n-1}(L) \stackrel{\text{def}}{=} (-1)^{n-1} \cdot (\text{coeff of } x \text{ in } \det(xI - L))$



$$= \sum_{i=1}^n \prod_{j \neq i} \lambda_j$$

$$= \sum_{\substack{S \subseteq V \\ |S|=n-1}} \det(L(S, S))$$

2017 (Schild): obs for sampling in time $m^{1+o(1)}$

2018 (Anari et al.): approx counts + sampling spanning forests

- Corollaries:
- poly-time algorithm for counting spanning trees
 - poly-time alg for sampling spanning trees (simple T w.p. prop to $w(T)$)

- Toss a biased coin to include edge e w.p. $\frac{\# \text{spanning trees in } G'}{\# \text{spanning trees in } G}$
- if don't include $e \rightarrow$ remove $e \rightarrow$ recurse on G' to set rest
 - if include $e \rightarrow$ contract e to a single vertex \rightarrow recurse on G'' to set rest

Proof: Write $L = \sum_{(a,b) \in E} w_{ab} (s_a - s_b)(s_a - s_b)^T = UTWU$

$= B^T B$

where $B = \begin{pmatrix} \vdots \\ \sqrt{w_{ab}}(s_a - s_b) \\ \vdots \end{pmatrix}$

$$\begin{aligned} \sigma_{n-1}(L) &= \sigma_{n-1}(B^T B) \\ &= \sigma_{n-1}(B B^T) \\ &= \sum_{\substack{S \subseteq E \\ |S|=n-1}} \det((B B^T)(S, S)) \end{aligned}$$

e-values of AB = e-values of BA

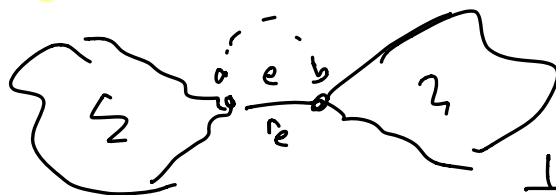
Recap: Effective Resistance

G undirected, weighted, connected graph.

View each edge as a resistor w/ resistance $r_e = \frac{1}{w_e}$

$$R_{\text{eff}}(a,b) = [v(a) - v(b) \text{ when } i_{\text{ext}} = \delta_a - \delta_b]$$

$$= (\delta_a - \delta_b)^T L^+ (\delta_a - \delta_b)$$



if removing e disconnects the graph, then

$$l_e = 1$$

$$\frac{1}{r_e} = w_e = \frac{1}{R_{\text{eff}}(e)}$$

Def: leverage score of edge e : $l_e = w_e R_{\text{eff}}(e)$

Thm: $\Pr[e \in T] = l_e$

T a random spanning tree chosen w.p. $\propto w(T)$

$$\text{Proof: } \Pr_T [e \in T] = \frac{\sum_{T: e \in T} w(T)}{\sum_T w(T)}$$

matrix-tree theorem

$$= \frac{\sum_{T: e \in T} \delta_{n-1}(L_T)}{\delta_{n-1}(L)}$$

$$\begin{aligned}
&= \sum_{\substack{S \subseteq E \\ |S|=n-1 \\ e \notin S}} \frac{\sigma_{n-1}(L_S)}{\sigma_{n-1}(L)} \\
&= \sum_{\substack{S \subseteq E \\ |S|=n-1 \\ e \notin S}} \sigma_{n-1}(L_S L^+) \\
&= \sum_{\substack{S \subseteq E \\ |S|=n-1 \\ e \notin S}} \sigma_{n-1} \left(\underbrace{B(S; \cdot)^T B(S; \cdot)}_A L^+ \right) \\
&= \sum_{\substack{S \subseteq E \\ |S|=n-1 \\ e \notin S}} \sigma_{n-1} \left(B(S; \cdot) L^+ B(S; \cdot)^T \right)
\end{aligned}$$

$$= \sum_{\substack{S \subseteq E \\ |S|=n-1 \\ e \notin S}} \sigma_{n-1}(\Gamma(S; S)) \quad \text{mxn}$$

for $\Gamma = B L^+ B^T$

$$\Gamma^2 = \Gamma$$

$$= \sum_{\substack{S \subseteq E \\ |S|=n-1 \\ e \notin S}} \sigma_{n-1} \left(\underbrace{\Gamma(S; \cdot)}_{\det} \underbrace{\Gamma(\cdot; S)}_{\det} \right)$$

det s
as volumes

$$= \sum_{\substack{S \subseteq E \\ |S|=n-1 \\ e \notin S}} \|v_e\|^2 \cdot \sigma_{n-2} \left(\underbrace{\Gamma \Pi_e \Gamma}_{\text{rank } n-2} (S, S) \right)$$

for $v_e = \Gamma(e; \cdot)$
 $\Pi_e =$ projection orthogonal to v_e

① symmetric matrix
nd $\Gamma^2 = \Gamma$

② symmetric matrix
nd all e-values
are 0 or 1

$$= \|v_e\|^2 \cdot \sigma_{n-2}(\Gamma \Pi_e \Gamma)$$

$$= l_e \cdot 1$$

$B^T B = I$

Γ a symmetric projection: $\Gamma = B L^T B^T$ has same e-values as $L^T B^T B = L^T L$
 $e = (a, b)$ $\text{Reff}(a, b)$ = projection onto \mathbb{R}^+

$$\Gamma(e, e) = \sqrt{w_{ab}} (\delta_a - \delta_b) L^T (\delta_a - \delta_b)^T \sqrt{w_{ab}} = \underline{le}$$

$$\|le\|^2 = \|\Gamma(e, \cdot)\|^2$$

$$\sigma_{n-1}(\Gamma(s; \cdot) \Gamma(\cdot; s)) = \det \left(\begin{matrix} \Gamma(s; \cdot) \\ \hline -\delta_{e_1} \\ -\delta_{e_2} \\ \vdots \\ -\delta_{e_{n-1}} \end{matrix} \right) \left(\begin{matrix} | & | & | \\ \delta_{e_1} & \delta_{e_2} & \dots & \delta_{e_{n-1}} \\ | & | & | \end{matrix} \right)$$

Π_{e_i} = proj orthogonal to δ_{e_i}

Π_e

project. orthogonal to a vector

$$\begin{aligned}
 &= \text{Vol}_{n-1} (\delta_{e_1}, \delta_{e_2}, \dots, \delta_{e_{n-1}})^2 \\
 &= (\|\delta_{e_1}\| \cdot \text{Vol}_{n-2} (\Pi_{e_1} \delta_{e_2}, \dots, \Pi_{e_1} \delta_{e_{n-1}}))^2 \\
 &= (\|\delta_{e_1}\| \cdot \det (\Gamma(s; \cdot) \Pi_{e_1} \Pi_{e_1} \Gamma(\cdot; s)))^2 \\
 &= (\|\delta_{e_1}\| \cdot \det (\Gamma(s; \cdot) \Pi_e \Gamma(\cdot; s)))^2
 \end{aligned}$$

σ_{n-2}

$$= \underline{I - \frac{vv^T}{\|v\|^2}}$$