

2025-Apr-2 For every weighted graph $G = (U, E, \omega)$ and $\varepsilon > 0$ $\exists H = (U, F, \mu)$ s.t.

$$(1-\varepsilon)G \leq H \leq (1+\varepsilon)G, |F| \leq \frac{4n \ln n}{\varepsilon^2} \text{ where } n = |U|, \text{ and } F \subseteq E.$$

G and H are similar in

sizes of boundaries of subsets
eigenvalues

inverses

effective resistances

Recall, $A \leq B$ iff $Bx - x^T Ax \leq x^T Bx$

Lemma For non-singular C

$$0 \leq A \leq B \Leftrightarrow C^T AC \leq C^T BC$$

Cor $A \leq B$ iff $B^{-1} \leq A^{-1}$

$$A \leq B \Leftrightarrow B^{-1/2} A B^{-1/2} \leq I \Leftrightarrow B^{1/2} A^{-1} B^{1/2} \geq I \Leftrightarrow A^{-1} \geq B^{-1}$$

For every edge $e \in E$, choose $p_e \in [0, 1]$. Then Matrix Chernoff Bounds (Tropp '12)

Include e in F with prob p_e . If include, set

$$\mu_e = \omega_e / p_e. \text{ Or, set } \mu_e = 0 \text{ if do not include}$$

$$\mathbb{E} \mu_e = p_e (\omega_e / p_e) = \omega_e \quad p_e = \alpha \omega_{a,b} \text{ Reff}(a, b)$$

$$\text{So, } \mathbb{E} L_H = \sum_{(a,b) \in E} (\mathbb{E} \mu_e) l_{a,b} = L_G$$

worry is deviations.

Let X_1, \dots, X_m be random psd matrices with $\|X_i\| \leq R$. $X = \sum X_i$,

$\mu_{\min} = \text{smallest eig of } EX$. $\mu_{\max} = \text{largest eig of } EX$.

$$\Pr[\lambda_{\min}(X) \leq (1-\varepsilon)\mu_{\min}] \leq n e^{-\frac{\varepsilon^2 \mu_{\min}}{2R}}$$

$$\Pr[\lambda_{\max}(X) \geq (1+\varepsilon)\mu_{\max}] \leq n e^{-\frac{\varepsilon^2 \mu_{\max}}{3R}}$$

Transformation

Let $L^{+1/2} = \left(L^+\right)^{1/2}$. Square root of pseudo-inverse

$$L_G^{+1/2} L_H L_G^{+1/2} = \sum X_{a,b} \text{ where}$$

$$(1-\varepsilon) L_G \leq L_H \leq (1+\varepsilon) L_G$$

↑ (Lemma 1, rish)

$$X_{a,b} = \begin{cases} \frac{w_{a,b}}{p_{a,b}} L_G^{+1/2} L_{a,b} L_G^{+1/2} & \text{with prob } p_{a,b} \\ 0 & \text{with prob } 1 - p_{a,b} \end{cases}$$

$$(1-\varepsilon) \Pi \leq L_G^{+1/2} L_H L_G^{+1/2} \leq (1+\varepsilon) \Pi$$

$$L_{a,b} = (\delta_a - \delta_b)(\delta_a - \delta_b)^T, \text{ so } \|L_G^{+1/2} L_{a,b} L_G^{+1/2}\| = \text{take min with } 1.$$

$$\text{where } \Pi = L_G^{+1/2} L_G L_G^{+1/2} = I - \frac{1}{n} I$$

$$(\delta_a - \delta_b)^T L_G^{+1} (\delta_a - \delta_b) = \text{Reff}(a,b) - p_{a,b} = \gamma w_{a,b} \text{Reff}(a,b)$$

$$\text{gives } \|X_{a,b}\| \in \{y, 0\}. \text{ So } R = y \gamma \text{ in Thm 1.}$$

$$\text{By Thm 1, prob is not an } \varepsilon\text{-approx} \leq 2n e^{-\frac{\varepsilon^2}{3R}}, \text{ small if } R \leq \varepsilon^2 / 3 \ln(2n)$$

$$\text{Set } R = \varepsilon^2 / 3.5 \ln(2n) = \gamma \varepsilon \quad \gamma = 3.5 \ln(2n) / \varepsilon^2$$

What if $p_{a,b} > 1$? Just include those edges and sample on rest

let $E_0 = \text{edges } p_{a,b} > 1, E_1 = \text{rest of edges.}$

$$\mu_{\max} < 1$$

$$\text{Want } \left(\lambda_{\max} \left(\sum_{(a,b) \in E_1} X_{a,b} \right) \right) \leq \mu_{\max} \left(\mathbb{E} \sum_{a,b \in E_1} X_{a,b} \right) + \varepsilon \quad \text{let } \hat{\varepsilon} = \varepsilon / \mu_{\max}, \text{ so Thm 2 gives}$$

$$\text{Prob } \text{fail} \leq n \exp \left(-\frac{\hat{\varepsilon}^2 \mu_{\max}}{3R} \right) = n \exp \left(-\frac{\varepsilon^2}{3R \mu_{\max}} \right) \leq n e^{-\frac{\varepsilon^2}{3R}}$$

$$\text{How many edges? } E \# \text{edges} = \sum_{a,b} P_{a,b} \leq \sum_{a,b} w_{a,b} \text{Reff}_G(a,b) \quad \delta = 3.5 \ln(2n) / \varepsilon^2$$

$$\sum_{a,b} w_{a,b} \text{Reff}_G(a,b) = n-1. \text{ So, } E \# \text{edges} \leq \frac{3.5 \ln n}{\varepsilon^2}.$$

$$\begin{aligned} \text{proof } \sum_{a,b} w_{a,b} (\delta_a - \delta_b)^T L^+ (\delta_a - \delta_b) &= \sum_{a,b} w_{a,b} \text{Tr}(L^+ (\delta_a - \delta_b)(\delta_a - \delta_b)^T) \\ &= \text{Tr}\left(\sum_{a,b} w_{a,b} L^+ L_{a,b}\right) = \text{Tr}\left(L^+ \sum_{a,b} w_{a,b} L_{a,b}\right) = \text{Tr}(L^+ L) = n-1. \end{aligned}$$

Can use Hoeffding to prove unlikely # edges $\Rightarrow \frac{4 \ln n}{\varepsilon^2}$

Starts, if time. Can quickly approximate $\text{Reff}_G(a,b)$ $\forall a,b \in E$.

Time $\sim m^{\log_2/\alpha^2}$ for α accuracy.

$$\text{Idea: } \text{Reff}(a,b) = (\delta_a - \delta_b)^T L^+ (\delta_a - \delta_b) = \|L^{+1/2}(\delta_a - \delta_b)\|$$

Johnson-Lindenstrauss: For any set of m vectors $x_1, \dots, x_m \in \mathbb{R}^n$

$$\text{If } R \text{ is } d \times n \text{ matrix of iid entries } \|Rx_i\|_2^2 \approx_{1-\alpha} \|x_i\|_2^2 \cdot \text{const} \quad d = \frac{\ln m}{\alpha^2}$$

So, just need $R L^{+1/2} \delta_a$ for each a . Can compute quickly ...

Mitchell - how got thesis

Batson - Yale undergrad took then.
