

Cheeger Inequalities

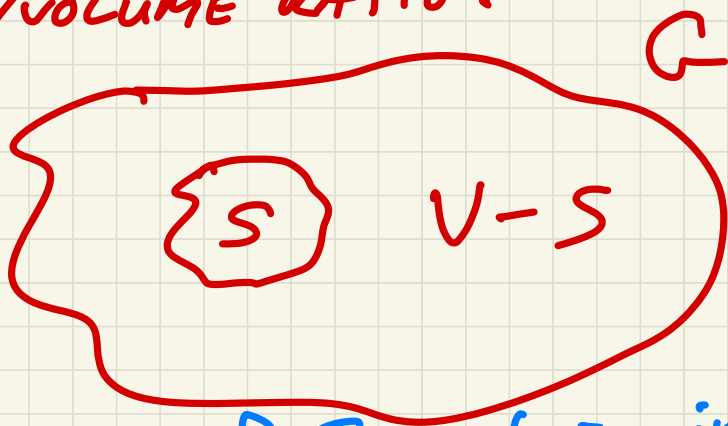
CS 292F
Lecture 6
April 21, 2021



Conductance $S \subseteq V = \frac{|\partial S|}{\min(d(S), d(V-S))} = \phi(S)$
 where $d(S) = \sum_{a \in S} (\text{degree of vertex } a)$

SURFACE/VOLUME RATIO:

In the picture,
 $|\partial S|$ scales
 like $\sqrt{d(S)}$



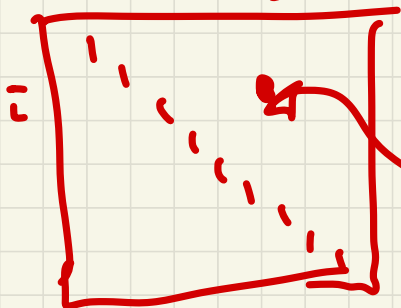
Conductance of G $\phi_G = \min_{S \subseteq V} \phi(S)$

smaller $\phi_G \Rightarrow$ easier to cut off a chunk.
 (clustering, partitioning)

Study Φ by "Normalized Laplacian"

$$N = N_G \stackrel{(\text{def})}{=} D^{-1/2} L_G D^{-1/2}$$

\uparrow \uparrow
 \mathcal{I}_i $\text{Diag}(\text{vertex degrees})$



$$\frac{-1}{\sqrt{d(i)d(j)}}$$

$$\frac{x^T N x}{x^T x} = \frac{y^T L y}{y^T D y} \quad \text{for } D^{1/2} y = x$$

so, the eigenvalues of N $Nx = \lambda x$
 are related to the generalized
 eigenproblem $L y = \lambda D y$

$N \not\subset 0$ (easy), evals $0 = \nu_1 \leq \nu_2 \leq \dots \leq \nu_n$

As with L , 0 is an eval (with multiplicity = #connected component of G). \nearrow of N

$$\nu_1 = 0$$

Eigenvector for $\nu_1 = 0$ is

$$\downarrow^{1/2} = \begin{pmatrix} \frac{\sqrt{d(1)}}{\sqrt{d(n)}} \\ \frac{\sqrt{d(2)}}{\sqrt{d(n)}} \\ \vdots \\ \frac{\sqrt{d(n)}}{\sqrt{d(n)}} \end{pmatrix}. \quad \text{Proof:}$$

$$N \downarrow^{1/2} = D^{-1/2} L D^{-1/2} \downarrow^{1/2} = D^{-1/2} L \underbrace{\downarrow^{1/2}}_0 = D^{-1/2} \underbrace{1}_{=0}$$

Recall we showed $\lambda_2/2 \leq \Theta_G$

THEOREM $\frac{\lambda_2}{2} \leq \Phi_G \leq \sqrt{2} \lambda_2$

Note: $\lambda_2 = \min_{x \perp d} \frac{x^T N x}{x^T x} = \min_{y \perp d} \frac{y^T L y}{y^T D y}$

Proof of $\lambda_2 \leq \Phi_G$:

use a test vector y :

For a set $S \subseteq V$, choose y to be constant on S and 0 on $V-S$.

To get $y \perp d$:

$y(i) = \begin{cases} d(v-s) & i \in S \\ -d(s) & i \in V-S \end{cases}$ so $y = \begin{matrix} d(v-s) \uparrow_S \\ -d(s) \downarrow_{V-S} \end{matrix}$

$y(i) = -d(s)$ for $i \in V - S$, $d(V - S)$ for $i \in S$

$$d^T y = \sum_i d(i) y(i) = \sum_{i \in S} d(i) d(V - S) + \sum_{i \in V - S} d(i) (-d(s))$$

$$= d(V - S) \sum_{i \in S} d(i) - d(s) \sum_{i \in V - S} d(i)$$

$$= d(V - S) d(s) - d(s) d(V - S)$$

$$= 0$$

so $y \perp d$

$$\frac{y^T L y}{y^T D y} :$$

$$\underline{y^T L y} = \sum_{i,j} (y(i) - y(j))^2 = \sum_{(i,j) \in E S} (-d(s) - d(v-s))^2$$

$$= \sum_{i,j \in E S} (d(v))^2 = \underline{d(v)^2 |E S|}$$

$$\underline{y^T D y} = \sum_i d(i) y(i)^2$$

$$= \sum_{i \in S} d(i) d(v-s)^2 + \sum_{i \notin S} d(i) d(s)^2$$

$$= d(s) (d(v-s))^2 + d(v-s) (d(s))^2$$

$$= d(s) d(v-s) (d(v-s) + d(s))$$

$$= \underline{d(s) d(v-s) d(v)}$$

$$\frac{y^T L y}{y^T D y} = \frac{d(v)^2 |E S|}{d(s) d(v-s) d(v)}$$

$$= \frac{d(v) |E S|}{d(s) d(v-s)}$$

$$\frac{y^T L y}{y^T D y} = \frac{d(v)^2 | \partial S |}{d(s) d(v-s) d(v)} = \frac{d(v) | \partial S |}{d(s) d(v-s)}$$

$$\text{Now } \nu_2 \leq \frac{y^T L y}{y^T D y} = \frac{d(v) | \partial S |}{d(s) d(v-s)}$$

$$\leq \frac{2 | \partial S |}{\min(d(s), d(v-s))} = 2 \phi(s)$$

This is true for all S ,

$$\text{so } \nu_2 \leq \min_S 2 \phi(s) = \Phi_G.$$

QED

Φ : ϕ (conductance)

Θ : Theta (isoper ratio)

Proof of $\Phi_G \leq \sqrt{2} \nu_2$

First part: from set $S \rightarrow$ test vector $y \perp d$.

This part: from vector $y \perp d \rightarrow$ set S

and show $\Phi(S) \leq \sqrt{2} \nu_2$

Let $y \perp d$.

Think of $y(i)$ labeling vertex i .

We'll take S to be

$$\underline{S} = \{a \in V : y(a) < \underline{t}\}$$

for some carefully
chosen t .

for all $y \perp d$:

we will show $\exists t$ such that

$$\Phi(S) \leq \sqrt{2 \frac{y^T L y}{y^T D y}}. \quad \text{Then } \Phi_G = \min_{y \perp d} \Phi(S) \leq \min_{y \perp d} \sqrt{2 \frac{y^T L y}{y^T D y}} = \sqrt{2} \nu_2$$

Start with $y \perp d$. Define $\rho = \frac{y^T L y}{y^T D y}$.

will show $\exists s \quad \phi(s) \leq \sqrt{2\rho}$

will show $\exists t : \phi(\{i: y(i) \leq t\}) \leq \sqrt{2\rho}$

① assume $y(1) \leq y(2) \leq \dots \leq y(n)$
(by renumbering the vertices)

② Take smallest k such that

$$\sum_{i=1}^k d(i) \geq \frac{d(v)}{2}$$

Then shift

$z = y + \alpha \mathbf{1}$ to make $z(k) = 0$

so

$$z(1) \leq z(2) \leq \dots \leq z(k) = 0 \leq \dots \leq z(n)$$

note:

$$z^T L z = (y + \alpha \mathbf{1})^T L (y + \alpha \mathbf{1}) = \\ y^T L y + 2\alpha y^T \underbrace{L \mathbf{1}}_0 + \alpha^2 \underbrace{\mathbf{1}^T L \mathbf{1}}_0 = y^T L y$$

$$z^T D z = (y + \alpha \mathbf{1})^T D (y + \alpha \mathbf{1}) \\ = y^T D y + \underbrace{2\alpha y^T D \mathbf{1}}_{2\alpha y^T d = 0} + \underbrace{\alpha^2 \mathbf{1}^T D \mathbf{1}}_{\alpha^2 d(V) \geq 0} \\ \geq y^T D y$$

$$\text{so } \frac{z^T L z}{z^T D z} \leq \frac{y^T L y}{y^T D y} = \rho$$

③ wlog, scale z so that

$$z(1)^2 + z(n)^2 = 1$$

note $z(1) < 0$ and $z(n) > 0$.

Want to show $\exists t \in [z(1), z(n)]$
 $z(1) \leq t \leq z(n)$

such that if $S = \{i : z(i) < t\}$

then $\frac{|2S|}{\min(d(S), d(V-S))} \leq \sqrt{2\rho}$.

(and then conclude

$$\underline{\Phi_G} = \min_S \Phi(S) \leq \min_y \sqrt{2\rho} = \underline{\sqrt{2\rho_2}}$$

• CHOOSE A CAREFUL PROBABILITY DISTRIBUTION ON the interval $[z(1), z(n)]$

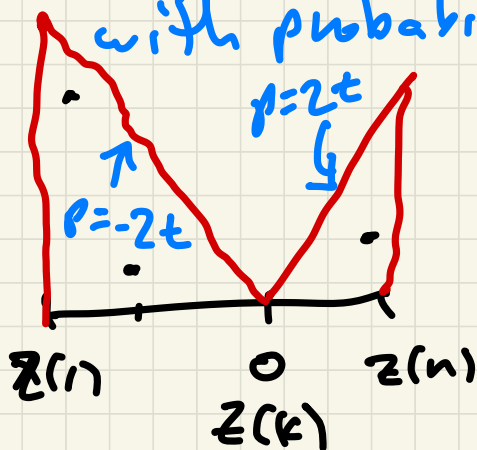
• SHOW
$$\mathbb{E}_t(|2S|) \leq \sqrt{2\rho} \mathbb{E}_t(\min(d(S), d(V-S)))$$

• Conclude $\exists t$ for which the inequality holds.

what probability distribution
on $z(i) \leq t \leq z(n)$?

(L. Trevisan 2011)

Choose t between $z(i)$ & $z(n)$
with probability density $2|t|$.



AND THE REST
IS JUST
ALGEBRA.

$$\Pr[a \leq t \leq b] = \int_a^b 2|t| dt$$

$$= (\text{sign } b) b^2 - (\text{sign } a) a^2$$

$$\int_{-z(i)}^{z(n)} 2|t| dt = z(i)^2 + z(n)^2 = 1$$

