

Graph Cuts & Eigenvalues

2-types of cuts

edge-cuts & vertex-cuts

Today edge-cuts

$G = (V, E, w)$ undirected weighted graph

Def $Cap(A, B) = \sum_{\substack{(a,b) \in E \\ a \in A, b \in B}} w(a, b)$

Cut measures

1) Quotient cuts or Isoperimetric number

$$\phi(G) = \min_{S \subseteq V} \frac{Cap(S, \bar{S})}{\min\{|S|, |\bar{S}|\}}$$

2) Sparsest Cut

$$\alpha(G) = \min_{S \subseteq V} \frac{\text{Cap}(S, \bar{S})}{|S| |\bar{S}|}$$

Note $\frac{n}{2} \alpha(G) \leq \beta(G) \leq n \alpha(G)$

3) Conductance

$$\Phi(G) = \min_{S \subseteq V} \frac{\text{Cap}(S, \bar{S})}{\min\{\text{Cap}(S), \text{Cap}(\bar{S})\}}$$

4) Sparest Cut

$$\lambda(G) = \min_{S \subseteq V} \frac{\text{Cap}(S, \bar{S})}{\text{Cap}(S) + \text{Cap}(\bar{S})}$$

another view

$$\text{Demand}(V_i, V_j) \equiv d_i \cdot d_j$$

$$\text{Demand}(S, \bar{S}) = \sum_{\substack{s \in S \\ s' \in \bar{S}}} d_s \cdot d_{s'}$$

$$= \left(\sum_{s \in S} d_s \right) \left(\sum_{s' \in \bar{S}} d_{s'} \right) = \text{Cap}(S) \cdot \text{Cap}(\bar{S})$$

more generally

$$\bar{Z}(G) = \min_{S \subseteq V} \frac{\text{Cap}(S, \bar{S})}{\text{Demand}(S, \bar{S})}$$

some demand fcn

eg product demand

$$\text{ie } \text{Demand}(v_i, v_j) = P_i \cdot P_j$$

note $\Phi(G)$ is $\approx P_L = \frac{d_i}{\sum d_j}$

Main Thms

$$L \equiv \text{Laplacian of } G \quad \lambda_2 = \lambda_2(L)$$

$$\text{Thm1} \quad \lambda_{2/2} \leq g(G) \leq \sqrt{2\Delta} \lambda_2$$

$$\Delta \equiv \max_i \{d_i\}$$

$$\mathcal{L} \equiv \text{Normalized Laplacian of } G \quad \lambda_2 = \lambda_2(\mathcal{L})$$

$$\text{ie } \mathcal{L} = D^{-1/2} L D^{-1/2}$$

$$\text{Thm2} \quad \lambda_{2/2} \leq \mathcal{E}(G) \leq \sqrt{2} \lambda_2$$

Claim $\lambda_2(L) \leq 2 \cdot g(G)$

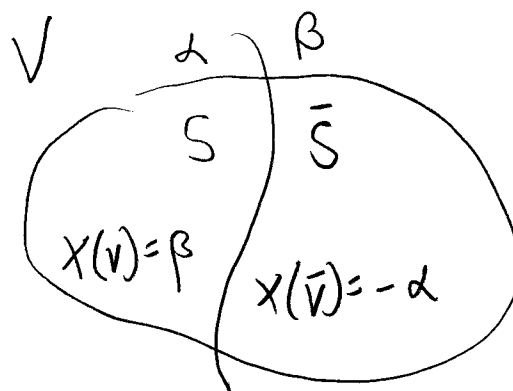
ie $\exists$ vector x st $x^T \mathbf{1} = 0$ & $\frac{x^T L x}{x^T x} \leq 2 \cdot g$

pf Let $S \subseteq V$ be set, $|S| \leq n/2$

$$\frac{\text{Cap}(S, \bar{S})}{|S|} = g$$

set $\alpha = |S|$ & $\beta = |\bar{S}|$

Def $x_i = \begin{cases} \beta & \text{if } i \in S \\ -\alpha & \text{o.w} \end{cases}$



$$x^T \mathbf{1} = \alpha \cdot \beta - \alpha \beta = 0$$

$$\frac{x^T L x}{x^T x} = \frac{\text{Cap}(S, \bar{S}) (\alpha + \beta)^2}{\alpha \beta^2 + \beta \alpha^2} = \frac{\text{Cap}(S, \bar{S}) n^2}{\alpha \cdot \beta \cdot n} = \left(\frac{\text{Cap}(S, \bar{S})}{\alpha} \right) \left(\frac{n}{\beta} \right)$$

$$\leq g \cdot 2$$

pf of upper bd for Thm 2.

Consider spring-mass version $LX = \lambda DX$

Given $X^T \begin{pmatrix} a_1 \\ \vdots \\ a_n \end{pmatrix} = 0$

We show $\Phi(G) \leq \sqrt{2 \frac{X^T L X}{X^T D X}}$ (*)

The proof is effective!

i.e. given X we find a "good" cut.

A "threshold cut"

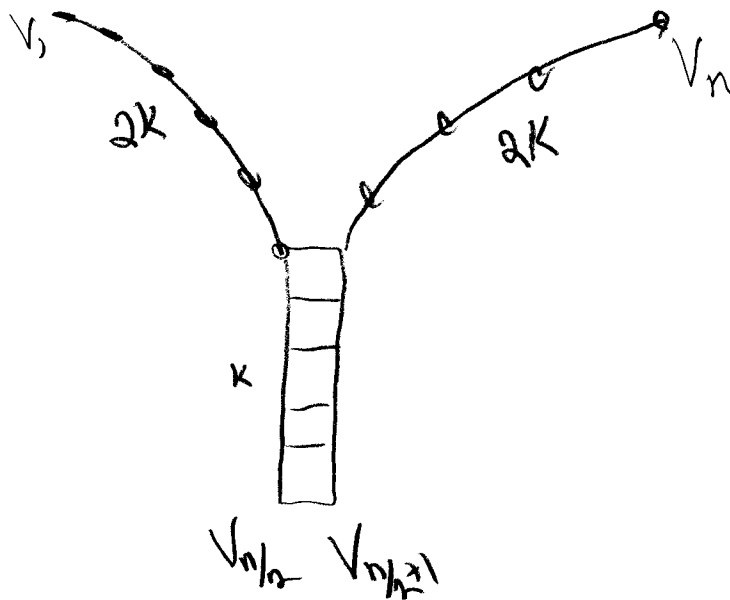
Suppose $f: V \rightarrow \mathbb{R}$

sort V so that $f(v_i) \geq f(v_{i+1})$

Def i th threshold cut $\equiv S_i = \{v_1, \dots, v_i\}$

Note $S_{n/2}$ may not be "good".

Consider Bug-Graph



Claim If $Lf = \lambda_2 f$ then f is odd with respect to its reflection.

The threshold cut for $S_{n/2}$ gives $\frac{K}{3K} \approx \frac{1}{3}$

$$g \approx \Phi \approx \frac{1}{2K} = \frac{3}{n}$$

Preliminaries $A \equiv \text{adj matrix}$

$$L = D - A \quad z^T L z = \sum_{i,j} w_{ij} (z_i - z_j)^2$$

The sum-Laplacian $P = D + A$

Claim $z^T P z = \sum_{i,j} w_{ij} (z_i + z_j)^2$

Fact $P = 2D - L$

Cauchy-Schwarz

$$\begin{aligned} (z^T L z)(z^T P z) &= \left(\sum w_{ij} (z_i - z_j)^2 \right) \left(\sum w_{ij} (z_i + z_j)^2 \right) \\ &= \left(\sum (\sqrt{w_{ij}} |z_i - z_j|)^2 \right) \left(\sum (\sqrt{w_{ij}} |z_i + z_j|)^2 \right) \\ &\geq \left(\sum w_{ij} |z_i^2 - z_j^2| \right)^2 \end{aligned}$$

The proof is st $z^T D \bar{1} = 0$ $z_1 \geq z_2 \geq \dots \geq z_n$

$$\Phi_i = \frac{\text{Cap}(S_i, \bar{S}_i)}{\min\{\text{Cap}(S_i), \text{Cap}(\bar{S}_i)\}}$$

To show $\exists i \quad \Phi_i \leq \sqrt{2 \frac{z^T L z}{z^T D z}}$

We modify z & G only decreasing $\frac{z^T L z}{z^T D z}$ s.t.

$$\beta = \arg \max_i \{ \text{Cap}(\bar{S}_i) \geq \text{Cap}(S_i) \}$$

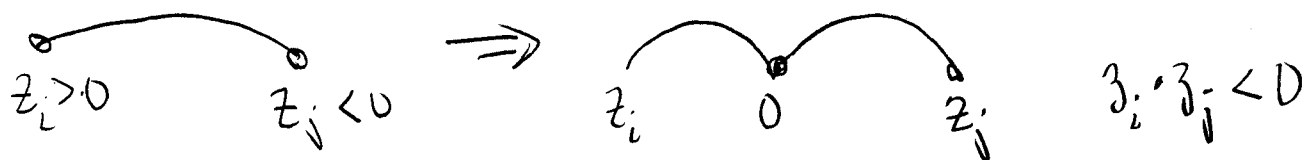
1) $z_\beta = 0$ (shifting z)

2) $\forall (i, j) \in E \quad z_i \cdot z_j \geq 0$ (add new point & set to zero)

1) Shifting z : $y = z + \alpha \bar{1}$

$$\frac{y^T L y}{y^T D y} = \frac{(z + \alpha \bar{1})^T L (z + \alpha \bar{1})}{(z + \alpha \bar{1})^T D (z + \alpha \bar{1})} = \frac{z^T L z}{z^T D z + 2\alpha z^T D \bar{1} + \alpha^2 \sum d_i} \leq \frac{z^T L z}{z^T D z}$$

pinching an edge at zero



Claim $\frac{y^T L' y}{y^T D' y} \leq \frac{z^T L z}{z^T D z}$

Note $y^T D' y = z^T D z$

To Show $y^T L' y \leq z^T L z$

Term $(z_i - z_j)^2$ replaced with $(z_i - 0)^2 + (0 - z_j)^2$

but $(z_i - z_j)^2 = z_i^2 - 2z_i z_j + z_j^2 > z_i^2 + z_j^2$

For sym-Laplacians

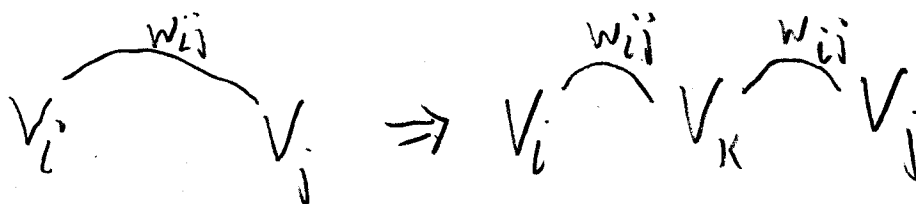
$$\frac{y^T P' y}{y^T D' y} = \frac{y^T (2D' - L') y}{y^T D' y} \leq \frac{y^T (2D') y}{y^T D' y} = 2$$

$$2 \frac{z^T L z}{z^T D z} \cdot (y^T D y)^2 \geq (y^T P' y)(y^T L' y) \geq \left(\sum_E w_{li} |y_i^2 - y_j^2| \right)^2$$

$$\sqrt{2 \frac{z^T L z}{z^T D z} (y^T D y)} \geq \sum_{\substack{E \\ i < j \leq \beta}} w_{li} (y_i^2 - y_j^2) + \sum_{\substack{E \\ j > i \geq \beta}} w_{li} (y_j^2 - y_i^2) \quad \neq$$

Lets assume $y_i > y_{i+1}$

WLOG Assume G is weighted path graph



new edge weights w_i from V_i to V_{i+1}

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$$\Phi_i = \frac{w_i}{\text{Cap}(V_i)} \quad \Phi^* = \min_{i \leq \beta} \Phi_i$$

$$\begin{aligned} \sum_{i=1}^{\beta} w_i (y_i^2 - y_{i+1}^2) &= \sum_{i=1}^{\beta} \Phi_i \text{Cap}(V_i) (y_i^2 - y_{i+1}^2) \\ &\geq \Phi^* \sum_{i=1}^{\beta} \text{Cap}(V_i) (y_i^2 - y_{i+1}^2) \\ &= \Phi^* \sum_{i=1}^{\beta} (\text{Cap}(V_i) - \text{Cap}(V_{i-1})) y_i^2 \\ &= \Phi^* \sum_{i=1}^{\beta} d_i y_i^2 \end{aligned}$$

by *

$$\sqrt{2 \frac{z^T L z}{z^T D z}} (y^T D y) \geq \Phi^* (y^T D y)$$

Thus

$$\sqrt{2 \frac{z^T L z}{z^T D z}} \geq \Phi^*$$