

# Partitioning and Isoperimetry

CS 292F

April 8, 2021

Lecture 3



$$0 = \lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$$

$$\lambda_1 = 0 \quad w_1 = \frac{1}{\sqrt{n}} \mathbf{1}$$

$$\lambda_2 = \min_{\mathbf{1}^T \mathbf{x} = 0} \frac{\mathbf{x}^T \mathbf{L} \mathbf{x}}{\mathbf{x}^T \mathbf{x}}$$

$$w_2 = \text{argmin}(\text{same})$$

$$\lambda_3 = \min_{\substack{\mathbf{1}^T \mathbf{x} = 0 \\ w_2^T \mathbf{x} = 0}} \frac{\mathbf{x}^T \mathbf{L} \mathbf{x}}{\mathbf{x}^T \mathbf{x}}$$

$$w_3 = \text{argmin}$$

$$w_2^T \mathbf{x} = 0$$

⋮

⋮

Note:  $\forall \mathbf{x}$  with  $\mathbf{1}^T \mathbf{x} = 0$ ,

$$\lambda_2 \leq \frac{\mathbf{x}^T \mathbf{L} \mathbf{x}}{\mathbf{x}^T \mathbf{x}}$$

# Graph Drawing

Draw in 1D.

Want to choose  $x(a)$  for  $a \in V$   
such that neighbors are close.

-  $\sum_{(a,b) \in E} (x(a) - x(b))^2$  small.  
I.E. make  $x^T L x$  as small as possible.

But  $x=0$  is boring. So  
require  $\|x\| = 1$

But  $x = \frac{1}{\sqrt{n}} \mathbf{1}$  is boring

So require  $\sum_a x(a) = 0 = \mathbf{1}^T x$

so  $x$  is argmin  $x^T L x$  Just  $w_2$ !  
 $\|x\| = 1$   
 $\mathbf{1}^T x = 0$

In 2D, put a at  $(x(a), y(a))$

$$\min \sum_{a, b \in E} \left\| \begin{pmatrix} x(a) \\ y(a) \end{pmatrix} - \begin{pmatrix} x(b) \\ y(b) \end{pmatrix} \right\|^2$$

Subject to

$$\|x\| = 1$$

$$\|y\| = 1$$

$$1^T x = 0$$

$$1^T y = 0$$

$$y^T x = 0$$

gives

$$x = w_2$$

$$y = w_3$$

$$\searrow \\ x^T L x + y^T L y$$

See Spielman  
section 3.2  
for  
details.

# GRAPH PARTITIONING

SPLIT VERTICES EQUALLY, WITH  
FEW EDGES BETWEEN PARTS  
(2 parts)

$\tau_{\min}$  is NP-hard

Define a "cut vector"  $x \in \mathbb{R}^V$

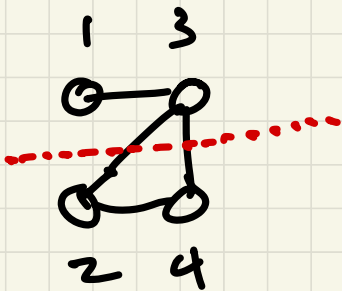
$$\text{as } x(a) = \pm 1$$

$$G = A \cup B$$

$$x = \mathbf{1}_A - \mathbf{1}_B$$

$$x^T L x = \sum_{(a,b) \in E} (x(a) - x(b))^2$$

$$= 4 \times (\# \text{ edges crossing cut})$$



$$x = \begin{pmatrix} 1 \\ -1 \\ -1 \\ -1 \end{pmatrix}$$

$$x^T \begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & 2 & -1 & -1 \\ -1 & -1 & 3 & -1 \\ 0 & -1 & -1 & 2 \end{bmatrix} x = 8$$

$L$

$$\min_x (x^T L x)$$

$$x(a) \in \pm 1$$

$$\sum_a x(a) = 0, \text{ i.e. } \mathbf{1}^T x = 0$$

NP-complete

$x$  is a corner of the hypercube in  $n$ -D.

$$\lambda_2 = \min_x (x^T L x) \text{ and } w_2 = \operatorname{argmin}_x$$

$$\|x\| = 1$$

$$\mathbf{1}^T x = 0$$

$x$  is a point on the sphere in  $n$ -D.

EASY

FIEDLER VECTOR

# ISOPERIMETRIC RATIOS

Graph  $G = (V, E)$

set  $S \subseteq V$

DEF Boundary  $\partial(S) = \{(a, b) \in E : a \in S, b \notin S\}$

DEF Isoperimetric Ratio

$$\Theta(S) = \frac{|\partial(S)| \text{ \#edges }}{|S| \text{ \#vtxs }}.$$

DEF

$$\Theta_G = \min_{|S| \leq \frac{n}{2}} \Theta(S)$$

THM For all  $S \subseteq V$ ,

$$\Theta(S) \geq \lambda_2(1-\sigma), \text{ where}$$

$$\sigma = |S|/n$$

THUS  $\Theta_G \geq \lambda_2/2$  "  $\lambda_2$  big  $\Rightarrow$  Graph well connected "

PROOF:  $\lambda_2 = \min_{x^T \mathbf{1} = 0} x^T L x / x^T x$

Take  $x = \mathbf{1}_S - \sigma \mathbf{1}$   $x^T \mathbf{1} = \mathbf{1}_S^T \mathbf{1} - \sigma \mathbf{1}^T \mathbf{1}$

$$x^T x = n(\sigma - \sigma^2) = |S|(1-\sigma) = |S| - \sigma n = 0$$

$$x^T L x = \sum_{(a,b) \in E} (x(a) - x(b))^2 = | \delta S |$$

$$\lambda_2 \leq \frac{x^T L x}{x^T x} = \frac{| \delta S |}{|S|(1-\sigma)} = \Theta(S) \frac{1}{(1-\sigma)}$$

$$\text{Thus } \lambda_2(1-\sigma) = \Theta(S)$$

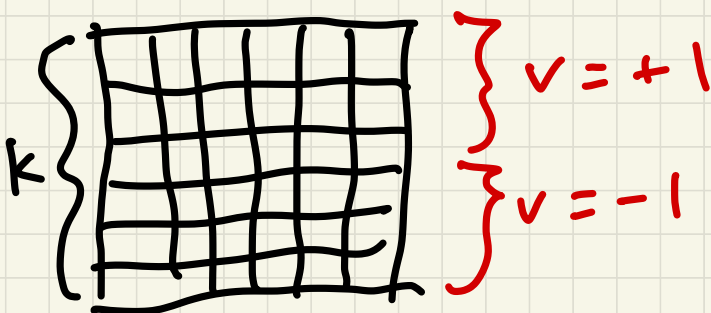


TEST VECTOR:  $v$  with  $\mathbf{1}^T v = 0$

$$\lambda_2 \leq \frac{v^T L v}{v^T v}$$

---

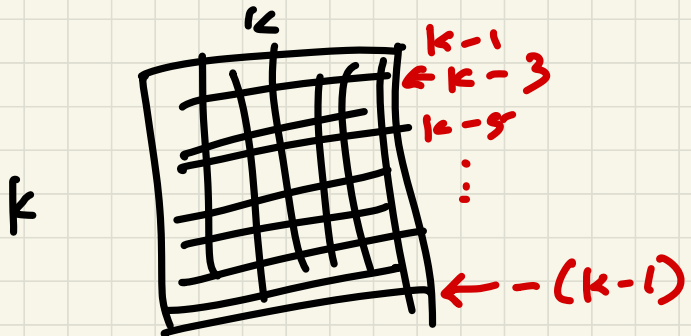
Grid graph:  $n = k^2$  vertices



$$v^T L v = 4k = 4\sqrt{n}$$

$$v^T v = n$$

$$\lambda_2 \leq \frac{v^T L v}{v^T v} = \frac{4}{\sqrt{n}} = O\left(\frac{1}{\sqrt{n}}\right)$$



$$n = k^2$$

$$-(k-1) \leq v(a) \leq k-1 \text{ for all } a \in V$$

$$\mathbf{1}^T \mathbf{V} = 0$$

$$\begin{aligned} \mathbf{V}^T \mathbf{V} &= ((k-1)^2 + (k-3)^2 + \dots + (-k+1)^2) \cdot k \\ &= \Theta(k^4) = \Theta(n^2) \end{aligned}$$

$$\mathbf{V}^T \mathbf{L} \mathbf{V} = \sum_{a,b \in E} (v(a) - v(b))^2$$

$$= 2^2 \cdot \Theta(n) = \Theta(n)$$

$$\lambda_2 \leq \frac{\mathbf{V}^T \mathbf{L} \mathbf{V}}{\mathbf{V}^T \mathbf{V}} = \frac{\Theta(n)}{\Theta(n^2)} = \Theta\left(\frac{1}{n}\right)$$

TIGHT.

BIG THEOREM:  $\lambda_2 = \Theta\left(\frac{1}{n}\right)$  for every planar graph of bounded degree.

rank  $\left(\frac{1}{n}\right)$  degree

