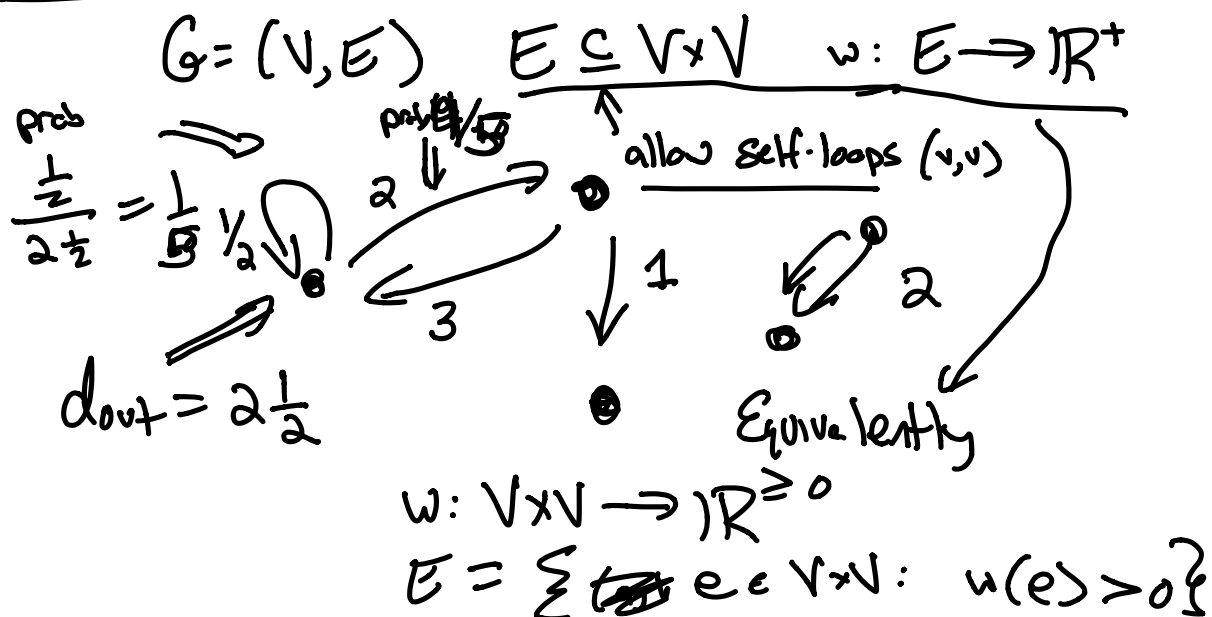


Announcements

- include pronouns in Zoom name
- mute self when not speaking
- videos on, don't record other students or share class recordings
- TFs: Chi-Ning Chau, Alec Sun, Santhoshini Velusamy, + TBD
- scribe
- use Zoom raise hand
- start recording
- live sync of notes
- PSO due Fri 9/11
- my OH Thurs 9-10am
- fill out section + OH poll ASAP
- use Piazza to find peer partners
- in case Zoom goes down, check Piazza

Graphs most general for us = weighted digraph



Special cases → Spielman + TODAY

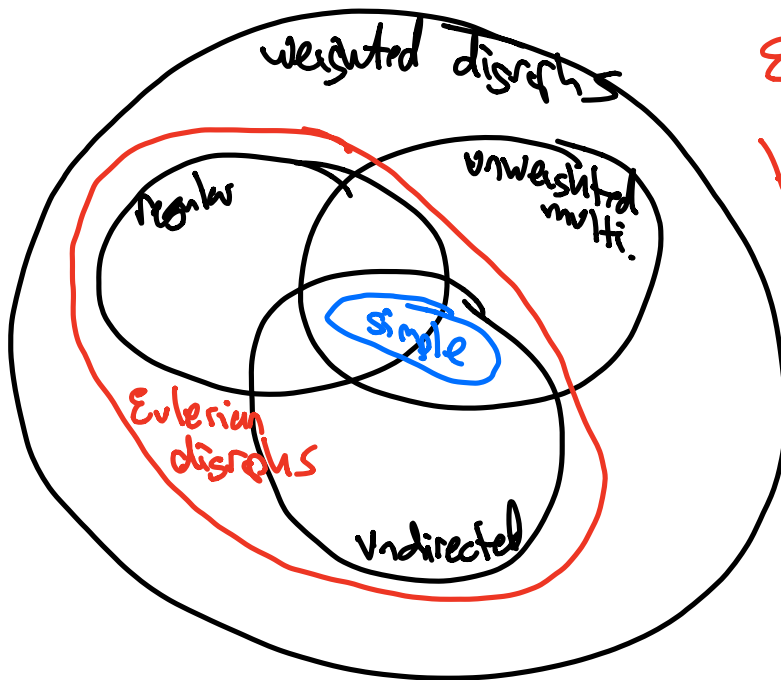
- undirected: $w(a,b) = w(b,a) \quad \forall a,b$

- unweighted multigraph: $w: V \times V \rightarrow \mathbb{N}$

- regular: $\exists d \quad w(a,b) = \text{multiplicity of } (a,b)$
 $\forall a \quad d_{\text{out}}(a) = d_{\text{in}}(a) = d$

$$d_{\text{out}}(a) \stackrel{\text{def}}{=} \sum_b w(a,b)$$

$$d_{\text{in}}(a) = \sum_b w(b,a)$$



Eulerian digraphs

$\forall a \quad d_{\text{in}}(a) = d_{\text{out}}(a)$

Simple graph

undirected
 unweighted
 no parallel edges
 no self-loops

Matrices M_G

- adjacency matrix $M(a, b) = w(b, a)$

- random-walk matrix $W = M \cdot D_{out}^{-1}$
diffusion

$$D_{out} = \begin{pmatrix} d_{out}(1) & & 0 \\ & \dots & \\ 0 & & d_{out}(n) \end{pmatrix}$$

$$W(a, b) = \frac{w(b, a)}{d_{out}(b)}$$

$$D_{in} = \dots$$

In Eulerian digraph^(*), $D_{in} = D_{out} = D$

(*) Laplacian: $L = D - M$

RW Laplacian: $L_{rw} = I - W \stackrel{(*)}{=} \underline{L \cdot D^{-1}}$

(*) Normalized Laplacian: $N = I - D^{-1/2} M D^{-1/2}$
 $= D^{-1/2} L D^{-1/2}$

Assume G Eulerian!

<u>Matrix</u>	<u>e-vector</u>	<u>e-value</u>	<u>symmetric if G undirected?</u>
L	$\vec{1}$	0	✓
$L_{rw} = I - W$	$\vec{d} = D\vec{1}$	0	not nec. when G irregular
W	\vec{d}	1	"
N	$\vec{d}^{1/2}$	0	✓
M	??		✓

Note: when G is d -regular all of these are equivalent to study

$$D = dI$$

$\Rightarrow \vec{1}$ is e-vector of $\overset{M}{*} = dI - L$
of e-~~vector~~^{value} d .

Spectral

Thm :

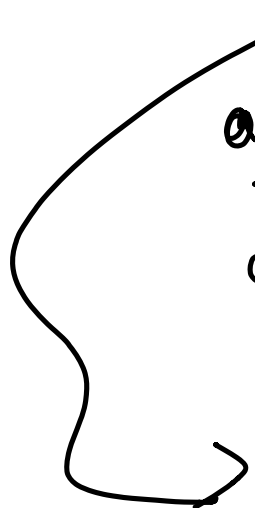
If M is a symmetric matrix,

then $M = V \Lambda V^T$ for an

orthogonal matrix V

$$\begin{pmatrix} VV^T = I \\ V^T V = I \\ \text{cols/rows ortho.} \\ \text{normal} \end{pmatrix}$$

and a diagonal matrix Λ .



$$M = \sum_{i=1}^n \lambda_i v_i v_i^T$$

diag of Λ cols of V

equivalently v_1, \dots, v_n are an
 orthonormal basis of e-vectors
 of M w/ e-values $\lambda_1, \dots, \lambda_n$

For asymmetric matrices, we have SVD

$$M = U \Sigma V^T \text{ for orthogonal } U, V \text{ nonneg diag } \Sigma$$

Pf outline: largest e-value of M

① Let $\mu_1 = \max \frac{x^T M x}{x^T x}$ } Rayleigh quotient
and v_1 - correspondy ~~unit~~ ^{unit} vector x

② Show that v_1 is an e-vector
of eigenvalue λ_1

(pf: calculus - gradient = 0)

③ Restrict to $(n-1)$ -dimensional
subspace

$$V_1^\perp = \{ w : w^T v_1 = 0 \}$$

check: $M V_1^\perp \subseteq V_1^\perp$

④ Apply induction on dimension
to $M|_{V_1^\perp} = M$ restricted
to V_1^\perp
($v_1^T M w = \lambda_1 v_1^T w$ ~~for all $w \in V_1^\perp$~~ = 0)



- Get orthonormal basis of e -vectors, an v_1
- Add v_1 to get basis of \mathbb{R}^n .

Computational Complexity of Linear Algebra

On $n \times n$ matrices ^{middle problem \rightarrow factoring} all of the following can be computed in $O(n^\omega)$ arithmetic ops:

$$(2 \leq \omega \leq 2.373)$$

- matrix mult.
- matrix inversion
- determinant
- characteristic polynomial
- eigendecomposition of symmetric matrices $\} \text{to obtain}$

• solving $Ax=b$ • SVD } arbitrary accuracy

For sparse matrices, prefer f (# nonzero entries)

Let G be undirected (for rest of today).
 We know $\vec{1}$ is an eigenvector of L of e-value 0.

Q: what can we say about the other e-values?

A: all nonnegative ($\Leftrightarrow L$ pos. semi-definite)

$$L = \sum_{(a,b) \in E} w_{a,b} L_{(a,b)} = \sum_{(a,b) \in E} w_{a,b} \begin{pmatrix} \delta_a - \delta_b \\ (\delta_a - \delta_b)^T \end{pmatrix}$$

\uparrow
 graph w/
 one edge

$$x^T L x = \sum_{(a,b)} w_{a,b} (x_a - x_b)^2 \geq 0$$

Let $0 = \lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$ be e-values

Thm: $\lambda_2 = 0 \iff G$ disconnected

PF: $\Leftarrow L = \begin{pmatrix} L_{G_1} & 0 \\ 0 & L_{G_2} \end{pmatrix}$

$$\text{e-values}(L) = \text{e-values}(L_{G_1}) \cup \text{e-values}(L_{G_2})$$

$$\Rightarrow \sum_{(a,b) \in E} w_{a,b} (x_a - x_b)^2 = 0$$

$$\Rightarrow x_a = x_b \text{ for every edge}$$

$$\Rightarrow x \text{ is a mult. } \vec{1}$$

\uparrow
G connected

$$\Rightarrow \text{only e-values of e-value } 0 \text{ are multiple of } \vec{1} = v_1$$

Intuition: larger $\lambda_2 \Rightarrow G$ more "highly connected".

Exercise: calculate $\lambda_2 \stackrel{d=n}{=} d$ for the complete graph on n vertices w/ self-loops (degree n)

• is this the largest possible value of λ_2/d ? ~~for a d-regular graph~~ ^{No-regular graph}

for a d -regular graph $\frac{\lambda_1}{d} = 2$

$M = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{pmatrix} = \mathbf{1}\mathbf{1}^T$

$L = \begin{pmatrix} n-1 & -1 & -1 & \dots & -1 \\ -1 & n-1 & -1 & \dots & -1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -1 & -1 & -1 & \dots & n-1 \end{pmatrix}$

$\mu_1 = n \quad \mu_2 = 0$

$\lambda_2 = d - \mu_2 = d = n$

$L = D - M = dI - M$

Q: characterization of λ_2 in terms

of quadratic form $x^T L x$?
 or Rayleigh quotient $\frac{x^T L x}{x^T x}$.

$$\lambda_2 = \min_{x \perp \vec{1}} \frac{x^T L x}{x^T x} =$$

$$x^T L x =$$

$$\sum_{a,b} w_{a,b} (x_a - x_b)^2$$

$$= \min_{S \subseteq \mathbb{R}^n} \max_{x \in S \setminus \{0\}} \frac{x^T L x}{x^T x}$$

S 2-dim subspace
 Courant-Fischer Thm

Hall's Graph Drawing Method

find $f: V \rightarrow \mathbb{R}^2$ minimizing

$$\sum_{(a,b) \in E} w_{a,b} \|f(a) - f(b)\|^2$$

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

s.t. $\|x\| = \|y\| = 1, x \perp \vec{1}, y \perp \vec{1}, x \perp y$

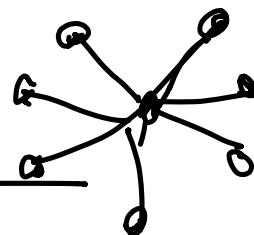
opt: $x = v_2$
 $y = v_3$

Wilf's Theorem:

Let $\mu_1 \geq \mu_2 \geq \dots \geq \mu_n$ be the eigenvalues of M . Then

G can be properly colored with

$\lfloor \mu_1 \rfloor + 1$ colors.



cf. - PSO: $d_{\max} + 1$ colors

- Fact: $\mu_1 \leq d_{\max}$

- $\mu_1 = d_{\max} \iff G$ regular

Proof of Wilf's Thm:

by induction on n

(1) find a vertex v of degree $\leq \lfloor \mu_1 \rfloor$

(2) color the rest of the graph

w/ $\lfloor \mu_1 \rfloor + 1$ colors

(3) color v w/ any color not used by a neighbor

(1) Lemma: in any graph G ,

$$\# d_{\text{avg}} \leq \mu_1$$

Pf: $\mu_1 = \max_x \frac{x^T M x}{x^T x} \geq \frac{\mathbf{1}^T M \mathbf{1}}{\mathbf{1}^T \mathbf{1}} = d_{\text{avg}}$

(2) by induction hypothesis +

Lemma: $G \setminus \{v\}$ has largest e-value $\leq \mu_1$

Pf: $\mu_1 = \max_{x \in \mathbb{R}^V} \frac{x^T M x}{x^T x}$

$$\geq \max_{y \in \mathbb{R}^{V-\{v\}}} \frac{\begin{pmatrix} y \\ 0 \end{pmatrix}^T M \begin{pmatrix} y \\ 0 \end{pmatrix}}{\begin{pmatrix} y \\ 0 \end{pmatrix}^T \begin{pmatrix} y \\ 0 \end{pmatrix}}$$

$$= \max_{y \in \mathbb{R}^{V-\{v\}}} \frac{y^T M_{G-\{v\}} y}{y^T y} \stackrel{\text{with constraint}}{\leq} \mu_1(G-\{v\})$$

Cauchy's Interlacing Thm:

A symmetric $n \times n$ matrix w/e-values $\alpha_1 \geq \dots \geq \alpha_n$

B principal $(n-1) \times (n-1)$ submatrix of -
w/e-values $\beta_1 \geq \dots \geq \beta_{n-1}$

Then $\alpha_1 \geq \beta_1 \geq \alpha_2 \geq \beta_2 \geq \dots \geq \beta_{n-1} \geq \alpha_n$
