

Spectral
9/29/09

Springs and Graph Laplacian's

Input: Graph of Springs (mattress)

$$G = (V, E, k)$$

$k_{ij} \equiv$ spring constant for edge E_{ij}

Consider only vertical displacements

$u_i \equiv$ displacement of V_i

Goal: Find solutions to Newton: $f = ma$

Find forces for displacement $u = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}$.

Set force on u_i by spring E_{ij}

$$-(u_i - u_j) k_{ij}$$

ie linear spring model.



$$L = L(G)$$

Force vector $-L u$

Let $m_i \equiv \text{mass of } V_i$ $M = \begin{pmatrix} m_1 & & 0 \\ & \ddots & \\ 0 & & m_n \end{pmatrix}$

$m_i > 0$

View u as a fcn of time $u(t) = \begin{pmatrix} u_1(t) \\ \vdots \\ u_n(t) \end{pmatrix}$

Acceleration: For $u_i(t) \equiv \frac{d^2 u_i}{dt^2}$

$$\frac{du}{dt} = \begin{pmatrix} \frac{du_1}{dt} \\ \vdots \\ \frac{du_n}{dt} \end{pmatrix} \quad \text{and} \quad a = \begin{pmatrix} a_1 \\ \vdots \\ a_n \end{pmatrix} = \frac{d^2 u}{dt^2}$$

Newton: $-L u = M \left(\frac{d^2 u}{dt^2} \right)$

Two ways to solve:

- 1) Solve for u given initial conditions
say $u(0)$ & $u'(0)$
- 2) Find steady state solutions

Do 2) using Guess and check.

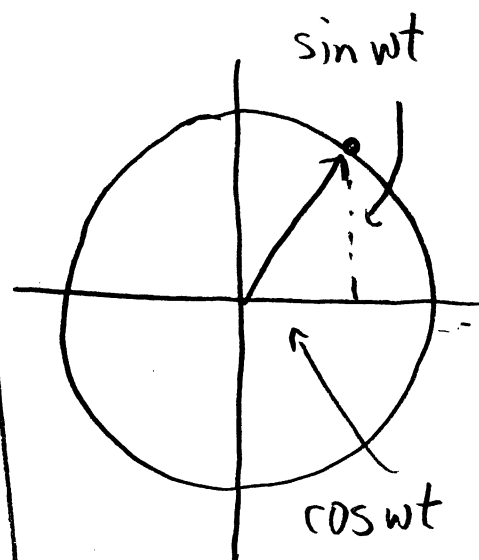
Recall: $e^{i\omega t} \equiv \cos \omega t + i \sin \omega t$

$$\frac{de^{i\omega t}}{dt} = i\omega e^{i\omega t} \equiv -\omega \sin \omega t + i\omega \cos \omega t$$

$$= i\omega (\cos \omega t + i \sin \omega t)$$

$$\frac{d^2 e^{i\omega t}}{dt^2} = \frac{d(i\omega e^{i\omega t})}{dt} = i^2 \omega^2 e^{i\omega t}$$

$$= -\omega^2 e^{i\omega t}$$



$$\frac{d \cos x}{dx} = -\sin x$$

$$\frac{d \sin x}{dx} = \cos x$$

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Guess: $u = e^{i\omega t} x$ some vector x .

$$\frac{d^2 u}{dt^2} = -\omega^2 e^{i\omega t} x$$

Check:

$$M(-\omega^2 e^{i\omega t} x) \stackrel{?}{=} -L(e^{i\omega t} x)$$

iff $\omega^2 Mx = Lx$ set $\lambda = \omega^2$

iff $Lx = \lambda Mx$ is (λ, x) is a generalized eigen-pair.

Claim: Eigenvalues of $Lx = \lambda Mx$ are real nonneg.

Change of variables $y = M^{1/2} x$ or $x = M^{-1/2} y$

$$L M^{-1/2} y = \lambda M M^{-1/2} y$$

$$\underbrace{M^{-1/2} L M^{-1/2}}_{L'} y = \lambda \cancel{M^{-1/2} M} M^{-1/2} y$$

L' is positive semidefinite

• We have found a space of dim n solutions.

eg $\omega_i = \sqrt{\lambda_i}$ then X_i then eigenvector

$$u(t) = \alpha_1 e^{i\omega_1 t} X_1 + \dots + \alpha_n e^{i\omega_n t} X_n$$

is a solution.

Consider setting mass = the weighted degree

Eigenspaces $LX = \lambda DX$ (*) $L = D - A$

change of variable $Y = DX$

(*) iff $(D - A)D^{-1}Y = \lambda Y$

$(I - AD^{-1})Y = \lambda Y$

$M = AD^{-1}$ transition matrix
for a random walk

iff $Y - MY = \lambda Y$

iff $-MY = (\lambda - 1)Y$ iff $MY = (1 - \lambda)Y$

If $0 = \lambda_1 < \lambda_2 \leq \dots \leq \lambda_n$ eigenvalues of (*)

then $0 \leq 1 - \lambda_n, \dots, 1 - \lambda_{n-1}, 0$ eigen of $MY = \lambda X$

Estimating λ_2 for Laplacian's

$$G = (V, E, w) \quad L = D - A = L(G)$$

eg

$$G = K_n$$

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

$$L \begin{pmatrix} u_1 \\ " \\ 1 \\ -1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} n \\ -n \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$L u_i = n u_i$$

$$u_i = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ -1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \Bigg\}_i$$

$$j = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}$$

eigenvector $j, u_1, \dots, u_{n-1}$

value $0, n, \dots, n$

We have full set

$$G = S_n = \begin{array}{c} x_1 \\ \circ \\ \diagup \quad \diagdown \\ x_0 \quad \circ \\ \diagdown \quad \diagup \\ \circ \quad x_n \end{array} \quad |V| = n+1$$

$$u_i = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ -1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \quad i = 1, \dots, n$$

$$u = \begin{pmatrix} \alpha \\ \beta \\ \vdots \\ \vdots \\ \beta \end{pmatrix}$$

$$\begin{pmatrix} n & -n \\ -n & n \end{pmatrix} x = \lambda \begin{pmatrix} 1 & 0 \\ 0 & n \end{pmatrix} x$$

$$\begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} x = \frac{\lambda}{n} \begin{pmatrix} 1 & 0 \\ 0 & n \end{pmatrix} x$$

$$G = L_n \text{ Line graph}$$

$$\text{Claim } \lambda_2 = \min_{\substack{V \perp \mathbf{1} \\ V \neq 0}} \frac{V^T L V}{V^T V}$$

$$(\geq) Lx = \lambda_2 x \Rightarrow x^T L x = \lambda_2 x^T x \text{ when } x \perp \mathbf{1} \text{ \& } x \neq 0$$

($\leq$) Spectral Thm

$$x = \left(-\frac{n}{2}, -\frac{n}{2}+1, \dots, +\frac{n}{2}\right)$$

$$x^T L x = \sum (x_i - x_{i+1})^2 = n-1$$

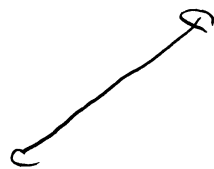
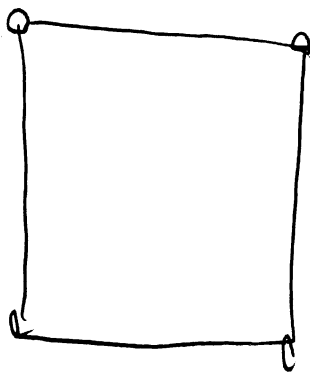
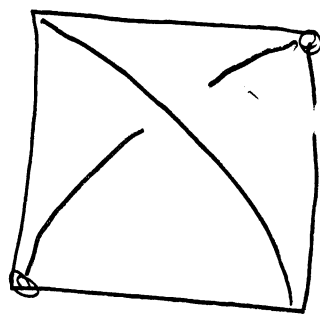
$$x^T x = 2 \sum_{i=1}^{n/2} i^2 \approx 2 \frac{(n/2)^3}{3} = \frac{n^3}{12}$$

$$\lambda_2 \leq \frac{n-1}{n^3/12} \leq \frac{12}{n^2}$$

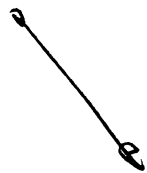
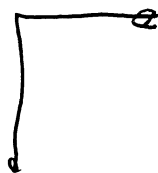
Def G (guest), H (host)

Path Embedding: $f: E_G \rightarrow \text{Paths in } H$

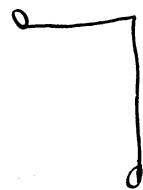
e.g. $G = K_n$ $H = C_4$



$\Rightarrow$



$\Rightarrow$



$$\text{Congestion} = \max_{e \in H} |P| \mid e \in P \mid \}$$

$$\text{Dilation} = \max_P \{ |P| \}$$

Thm $f: K_n \xrightarrow{\text{path}} G$ with congestion c & dilation d

then $\frac{n}{c \cdot d} \leq \lambda_2(G)$

Pf $A = L(G)$ $B = L(K_n)$

P_{ij} = Path from i to j

$L_{ij} = L(P_{ij})$

$Ax = \lambda_2 x$

$A \subseteq \frac{\sum L_{ij}}{c \cdot d}$

$\frac{\lambda_2}{n} = \frac{x^T A x}{x^T B x} =$

$\leq \frac{(x^T \sum L_{ij} x) / c \cdot d}{x^T \sum E_{ij} x} = \frac{1}{c \cdot d} \frac{\sum x^T L_{ij} x}{\sum x^T E_{ij} x}$

$\leq \text{Max} \frac{x^T L_{ij} x}{x^T E_{ij} x}$