

# Spectral Methods (aka Algebraic Graph Theory, see Chapter 14 in [BE] “Network analysis”)

Three main objects of interest: adjacency matrix, Laplacian, and normalized Laplacian

What their spectrum (all eigenvalues, including algebraic multiplicity) can tell about network statistics, existence of subgraphs, classification, etc.?

Let  $M \in \mathbb{C}^{n \times n}$ . A non-zero vector  $x \in \mathbb{C}^n$  is an eigenvector of  $M$  with corresponding eigenvalue  $\lambda \in \mathbb{C}$  if

$$Mx = \lambda x$$

The solution exists iff  $\text{rank}(M - \lambda I) < n$  iff  $\det(M - \lambda I) = 0$ , i.e., the eigenvalues are roots of  $\det(M - \lambda I) = 0$ .

If  $Q$  is non-singular then  $M$  and  $Q^{-1}MQ$  have the same eigenvalues.

• If  $M \in \mathbb{R}^{n \times n}$  and  $M = M^T$  then  $\exists$  non-singular  $Q$  s.t.  $Q^{-1} = Q^T$  and  $M' = Q^{-1}MQ$  has diagonal form. Eigenvectors of  $M'$  are  $e_i$  and

$$\det(M - \lambda I) = \det(M' - \lambda I) = \prod_i (\lambda_i - \lambda)$$

diagonal entry of  $M'$

One can infer  $\text{tr}(M) = \sum_{i=1}^n \lambda_i$  (check at home).

• If  $v_i = Qe_i$  then  $Mv_i = \lambda_i v_i$  and  $v_i^T v_j = e_i^T e_j$ .

orthonormal eigenvectors of  $M$

**Theorem 1.** Let  $M \in \mathbb{R}^{n \times n}$  and  $M = M^T$ , then

1.  $M$  has real eigenvalues  $\lambda_1 \leq \dots \leq \lambda_n$  and  $n$  orthonormal eigenvectors
2. multiplicity of  $\lambda_i$  as an eigenvalue = multiplicity of  $\lambda_i$  as a root of the characteristic polynomial  $\det(M - \lambda I)$  = cardinality of a maximum linearly independent set of eigenvectors corresponding to  $\lambda_i$
3.  $\exists Q$  with  $Q^T = Q^{-1}$  such that  $QMQ^{-1} = \text{diag}(\lambda_1, \dots, \lambda_n)$
4.  $\det(M) = \prod_i \lambda_i$  and  $\text{tr}(M) = \sum_i \lambda_i$

**Theorem 2.** Let  $G$  be a graph, and  $A$  its adj matrix with ordered eigenvalues  $\lambda_i$ , and  $\Delta$  is a max degree of  $G$  then

1.  $\lambda_n \leq \Delta$
2.  $G = G_1 \cup G_2 \implies \text{spec}(G) = \text{spec}(G_1) \cup \text{spec}(G_2)$
3.  $G$  is bipartite  $\implies$  (if  $\lambda \in \text{spec}(G)$  then  $-\lambda \in \text{spec}(G)$ )
4.  $G$  is simple cycle  $\implies \text{spec}(G) = \{2 \cos(\frac{2\pi k}{n}) \mid k \in \{1, \dots, n\}\}$
5.  $G = K_{n_1, n_2} \implies \lambda_1 = -\sqrt{n_1 n_2}, \lambda_2 = \dots = \lambda_{n-1} = 0$ , and  $\lambda_n = \sqrt{n_1 n_2}$
6.  $G = K_{n_1} \implies \lambda_1 = \dots = \lambda_{n-1}, \lambda_n = n - 1$

### Theorem 3.

1.  $\sum_{i=1}^n \lambda_i = \text{number of loops in } G$
2.  $\sum_{i=1}^n \lambda_i^2 = 2 \times \text{number of edges in } G$
3.  $\sum_{i=1}^n \lambda_i^3 = 6 \times \text{number of triangles in } G$

**Homework: Prove any 2 out of 3 in Theorem 3 (submit by 3/6/2014)**

Laplacian matrix  $L = D - A$

Incidence matrix  $B = (b_{i,e}) = \begin{cases} 1 & i \text{ is the head of } e \\ -1 & i \text{ is the tail of } e \\ 0 & \text{otherwise} \end{cases}$

For any  $x \in \mathbb{C}^n$   $x^T L x = x^T B B^T x = \sum_{ij \in E} (x_i - x_j)^2$

A graph  $G$  consists of  $k$  connected components if and only if  $\lambda_1(L) = \dots = \lambda_k(L) = 0$  and  $\lambda_{k+1}(L) > 0$ .

**Trees and Laplacian:** for every  $i \in \{1, \dots, n\}$  the number of spanning trees in  $G$  is equal to  $|\det(L_i)|$ , where  $L_i$  is obtained from the Laplacian  $L$  by deleting row  $i$  and column  $i$ . Moreover, the number of spanning trees is equal to  $\frac{1}{n} \prod_{i \geq 2} \lambda_i(L)$ .

Normalized Laplacian  $\mathcal{L} = D^{-1/2}LD^{-1/2}$ , i.e.,

$$(\mathcal{L}_{ij}) = \begin{cases} 1 & i = j \text{ and } d(i) > 0 \\ -1/\sqrt{d(i)d(j)} & ij \in E \\ 0 & \text{otherwise} \end{cases}$$

There exists some nonzero function  $w$  on nodes

$\lambda$  is an eigenvalue of  $\mathcal{L}$ , i.e.,  $\lambda w(i) = \frac{1}{\sqrt{d(i)}} \sum_{j \in N(i)} \left( \frac{w(i)}{\sqrt{d(i)}} - \frac{w(j)}{\sqrt{d(j)}} \right)$

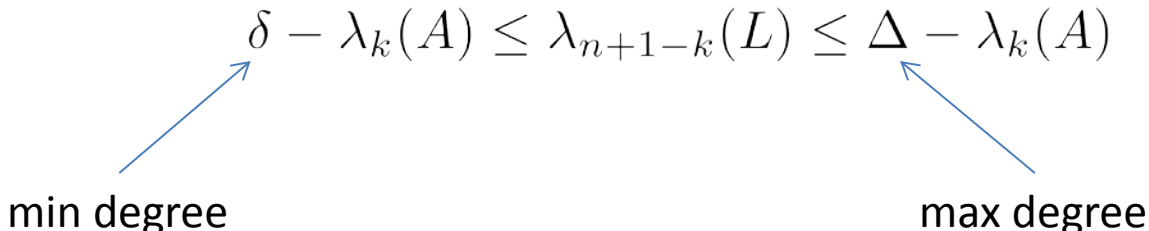
- $\lambda_1(\mathcal{L}) = 0$  and  $\lambda_n(\mathcal{L}) \leq 2$
- $G$  is bipartite iff for every  $\lambda(\mathcal{L})$  its "complement"  $2 - \lambda(\mathcal{L}) \in \text{spectra}(\mathcal{L})$
- $\lambda_i(\mathcal{L}) = 0, i \in [1..k] \implies G$  has  $k$  connected components

For  $d$ -regular graphs, the three spectra are equivalent

- $\text{spectra}(A) = \{\lambda_1, \dots, \lambda_n\}$
- $\text{spectra}(L) = \{d - \lambda_n, \dots, d - \lambda_1\}$
- $\text{spectra}(\mathcal{L}) = \{1 - \lambda_n/d, \dots, 1 - \lambda_1/d\}$

**Theorem:**  $\forall k$   $\lambda_k(A)$   $k$ th smallest eigenvalue of  $A$  and  $\lambda_{n-k+1}(L)$   $k$ th largest eigenvalue of  $L$

$$\delta - \lambda_k(A) \leq \lambda_{n+1-k}(L) \leq \Delta - \lambda_k(A)$$

  
min degree max degree

For a nonzero  $x \in \mathbb{R}^n$  and  $M \in \mathbb{R}^{n \times n}$  the Rayleigh quotient is defined

$$R(x) = \frac{x^T M x}{x^T x}$$

**Courant-Fischer Theorem.** Let  $M \in \mathbb{R}^{n \times n}$  be symmetric with eigenvalues  $\lambda_0 \leq \dots \leq \lambda_{n-1}$ . Let  $X^k$  be a  $k$ -dim subspace of  $\mathbb{R}^n$  and  $x \perp X^k$ . Then

$$\lambda_i = \min_{X^{n-i-1}} \left( \max_{x \perp X^{n-i-1}, x \neq 0} R(x) \right) = \max_{X^i} \left( \min_{x \perp X^i, x \neq 0} R(x) \right)$$

**Fiedler Theorem.**

$$\lambda_2(L) = n \min_{x \in \mathbb{R}^n} \left( \frac{\sum_{ij \in E} (x_i - x_j)^2}{\sum_{ij \in \binom{V}{2}} (x_i - x_j)^2} \right) \text{ same for } \lambda_n \text{ and } \max$$

A symmetric minor of  $A$  is a submatrix  $B$  obtained by deleting some rows and the corresponding columns.

**Theorem (Interlacing eigenvalues).** Let  $A \in \mathbb{R}^{n \times n}$  be a symmetric matrix with eigenvalues  $\lambda_1 \leq \dots \leq \lambda_n$ . Let  $B \in \mathbb{R}^{(n-k) \times (n-k)}$  be a symmetric minor of  $A$  with eigenvalues  $\mu_1 \leq \dots \leq \mu_{n-k}$ . Then

$$\lambda_i \leq \mu_i \leq \lambda_{i+k}.$$

**Corollary.** Let  $G$  and  $H$  be two graphs with eigenvalues  $\lambda_1 \leq \dots \leq \lambda_n$  and  $\mu_1 \leq \dots \leq \mu_m$  respectively. If  $\mu_1 < \lambda_1$  or  $\lambda_n < \mu_n$ , then  $H$  does not occur as an induced subgraph of  $G$ .

graph class	spectrum( $A$ )	spectrum( $L$ )	spectrum( $\mathcal{L}$ )
simple path $G = P_n$	$2 \cos\left(\frac{\pi k}{n+1}\right),$ $k \in \{1, \dots, n\}$	$2 - 2 \cos\left(\frac{\pi(k-1)}{n}\right),$ $k \in \{1, \dots, n\}$	$1 - \cos\left(\frac{\pi(k-1)}{n-1}\right),$ $k \in \{1, \dots, n\}$
simple cycle $G = C_n$	$2 \cos\left(\frac{2\pi k}{n}\right),$ $k \in \{1, \dots, n\}$	$2 - 2 \cos\left(\frac{2\pi k}{n}\right),$ $k \in \{1, \dots, n\}$	$1 - \cos\left(\frac{2\pi k}{n}\right),$ $k \in \{1, \dots, n\}$
star $G = K_{1,n}$	$-\sqrt{n}, \sqrt{n},$ $0$ ( $n - 2$ times)	$0, n,$ $1$ ( $n - 2$ times)	$0, 2,$ $1$ ( $n - 2$ times)
$G = K_{n_1, n_2}$	$-\sqrt{n_1 n_2}, \sqrt{n_1 n_2},$ $0$ ( $n - 2$ times)	$0, n_1$ ( $n_2 - 1$ times) $n_2$ ( $n_1 - 1$ times), $n$	$0, 2$ $1$ ( $n - 2$ times)
$G = K_n$	$1, -1$ ( $n - 1$ times)	$0, n$ ( $n - 1$ times)	$0, \frac{n}{n-1}$ ( $n - 1$ times)

# Computing Part of the Spectrum, Lanczos Algorithm

1. *Initialization*: Choose the number of steps  $k$ , the desired number of eigenvalues  $r$  and an initial vector  $x_1$ ; let  $\beta_0 := x_1^\top x_1$ ,  $x_1 := x_1/\beta_0$

2. *Lanczos steps*:

**for**  $i = 1$  *to*  $k$  **do**

- (i)  $y := Mx_i$
- (ii)  $\alpha_i := x_i^\top y$
- (iii)  $x_{i+1} := y - \alpha_i x_i - \beta_{i-1} x_{i-1}$
- (iv)  $\beta_i := x_{i+1}^\top x_{i+1}$
- (v)  $x_{i+1} := x_{i+1}/\beta_i$ ;

Set  $X_i := \text{Mat}(x_1, \dots, x_i)$

3. *Eigenvalue computation*: Compute the eigenvalues of  $T := X_i^\top M X_i$ .

4. *Convergence test and restart*: If the first  $r$  columns of  $T$  satisfy the convergence criteria then accept the corresponding eigenvalues and stop. Otherwise restart with a suitable new  $x_1$ .

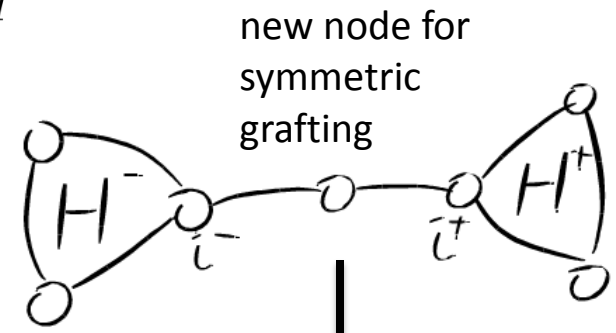
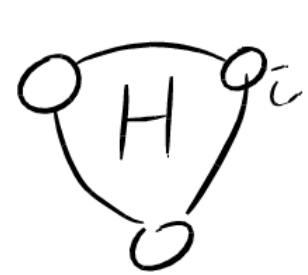
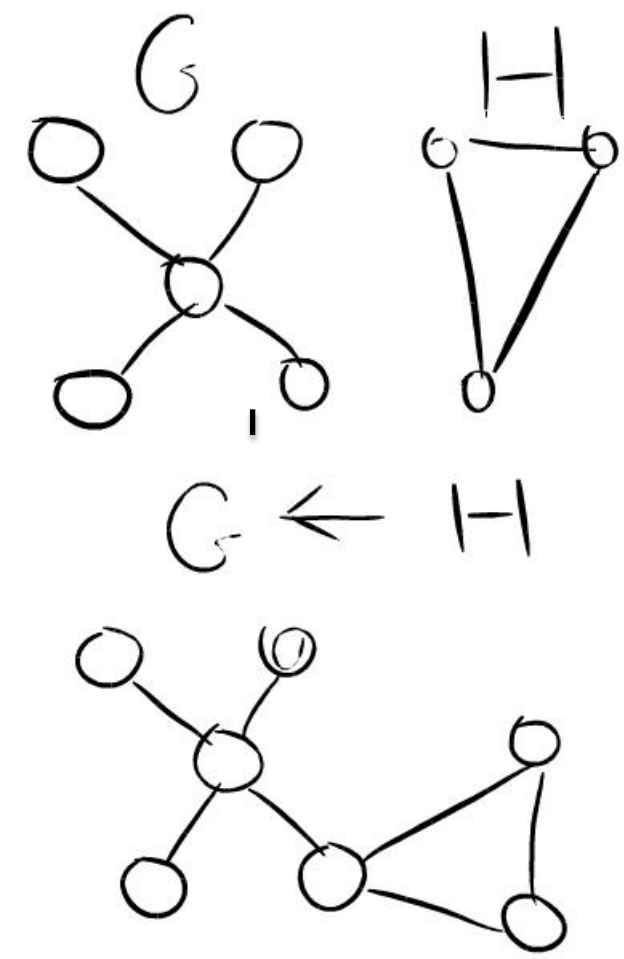


# Grafting

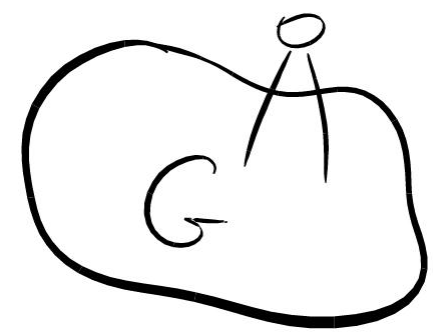
**Q: How to modify G by adding to it eigenvalues of H? What if we want to add  $\lambda$  to G?**

$(\lambda, x)$  is eigenpair of  $H$ ,  $\exists i \in V_H$   $x_i = 0$

$(\lambda, x)$  is eigenpair of  $H$



$\lambda \in \text{spectra}(G')$

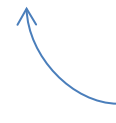


# Eigenvalues and Global Properties

- $\frac{1}{n} \sum_{i \in V} d(i) \leq \lambda_n(A)$

- $\left\lceil \frac{4}{n\lambda_2(L)} \right\rceil \leq \text{diam}(G) \leq 2 \left\lceil \frac{\cosh^{-1}(n-1)}{\cosh^{-1}\left(\frac{\lambda_n(L)+\lambda_2(L)}{\lambda_n(L)-\lambda_2(L)}\right)} \right\rceil + 1$

- $\frac{1}{n-1} \left( \frac{2}{\lambda_2(L)} + \frac{n-2}{2} \right) \leq \bar{\rho}(G) \leq \frac{n}{n-1} \left[ \frac{\Delta + \lambda_2(L)}{4\lambda_2(L)} \ln(n-1) \right]$


 mean distance in  $G$

- Isoperimetric number  $i(G) = \min \left\{ \frac{|cut(X,Y)|}{\min\{|X|,|Y|\}}; X \subset V, X \neq \emptyset, Y = V \setminus X \right\}$

$$i(G) \geq \min \left\{ 1, \frac{\lambda_2(L)\lambda_n(L)}{2(\lambda_n(L) + \lambda_2(L) - 2)} \right\} \quad i(G) \leq \sqrt{\lambda_2(L)(2\Delta - \lambda_2(L))}$$

- Expansion  $c_V := \min \left\{ \frac{|N(S) \setminus S|}{|S|}; S \subseteq V, |S| \leq \frac{n}{2} \right\} \quad \frac{\lambda_2(L)}{\frac{\Delta}{2} + \lambda_2(L)} \leq c_V = \mathcal{O}(\sqrt{\lambda_2(L)})$

... chromatic number, minimum independent set, ...