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Wednesday, April 8

Some remarks about the final exam:

Because the exam will be a take-home exam, and you will have four days to write it, the questions will require a bit more thought and creativity than the more routine questions you would expect in a 3 hour in-class exam.

In preparing for it, you should strive for a broad understanding of the concepts, in addition to going over past assignments.

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On Monday's lecture, we went over a few applications of the spectral theorem:

#1. Existence and uniqueness of positive definite square roots of positive definite self-adjoint operators.

#2. Polar decomposition:  $T = UP$ ,  
 $U$  unitary,  $P$  positive self-adjoint.

#3. Fourier analysis on finite groups

$$L^2(G) = \bigoplus_{\phi \in \hat{G} = \text{Hom}(G, \mathbb{C}^\times)} \mathbb{C} \cdot \phi \quad (\text{orthogonal direct sum.})$$

$$L^2(G) \cong L^2(\hat{G}) \quad \text{as inner product spaces.}$$

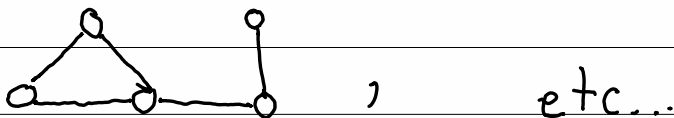
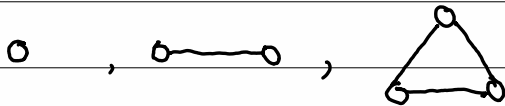
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## Application #4

## Spectral Graph Theory.

Definition: A graph  $G$  is a finite set  $V(G)$  of vertices together with a set  $E \subseteq V(G)^2$  of "edges" satisfying  $(v_1, v_2) \in E \Leftrightarrow (v_2, v_1) \in E$

Examples



, etc...

Graphs are the mathematical structure that can be used to

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Model networks, which are ubiquitous in many applications (the internet, distribution or communication networks, modelling the spread of infectious diseases, etc...)

Two vertices  $v, w \in V(G)$  are said to be adjacent if  $(v, w) \in E$

We write  $v \sim w$  if  $v$  and  $w$  are adjacent.

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## Adjacency operators

One can associate to a graph

$G$  the inner product space

$L^2(V) = \mathbb{C}$ -valued functions on the vertices,

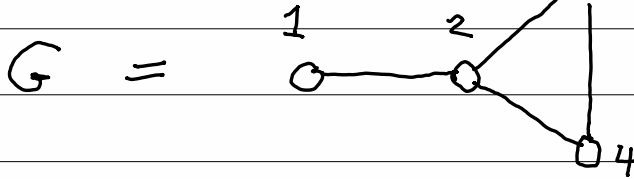
together with the Adjacency operator

$$A: L^2(V) \longrightarrow L^2(V)$$
$$f \longmapsto Af(v) = \sum_{w \sim v} f(w)$$

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Example

$$V(G) = \{1, 2, 3, 4\}$$



$(\delta_1, \delta_2, \delta_3, \delta_4) =$  orthonormal basis  
of  $L^2(V)$

In this basis,  $A$  is represented  
by the matrix

$$M = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{pmatrix}$$

Note:  $M = M^t$ .

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Proposition: The adjacency operator is self-adjoint.

$$\begin{aligned}\text{Proof: } \langle Af, g \rangle &= \sum_{r \in V(G)} Af(r) \overline{g(r)} \\ &= \sum_{r \in V(G)} \sum_{w \sim r} f(w) \overline{g(r)} \\ &= \sum_{(r, w) \in E} f(w) \overline{g(r)} = \sum_{w \in V(G)} \sum_{v \sim w} f(w) \overline{g(v)} \\ &= \sum_{w \in V(G)} f(w) \overline{(Ag)(w)} = \langle f, Ag \rangle.\end{aligned}$$

Corollary: The space  $L^2(V(G))$ .

has an orthonormal basis of eigenvectors for the adjacency operator.

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$$\text{Spectrum}(A) = \{ \lambda_1, \lambda_2, \dots, \lambda_t \}$$

$$\lambda_1 \geq \lambda_2 \geq \lambda_3 \geq \dots \geq \lambda_t \in \mathbb{R}$$

Main questions of spectral graph theory

① What geometric and combinatorial properties of  $G = (V(G), E)$  can be read off from the spectrum of  $A$ ?

② What is the meaning of the eigenfunctions for  $A$ ?



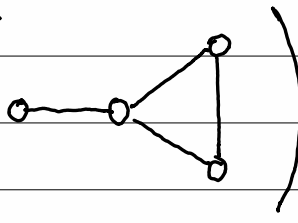
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## A sample result

Definition: The diameter of a graph is the maximum distance

between vertices.  $\delta(G) :=$  diameter of  $G$

E.g.,

$$\delta \left( \begin{array}{c} \circ \\ \circ - \circ \\ \circ \end{array} \right) = 2$$


Theorem:  $\# \text{ spectrum}(A) \geq \delta + 1$

Proof: For  $v, w \in V(G)$

$\langle A^k v, w \rangle = \#$  of paths of length  $k$  from  $v$  to  $w$ .

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In particular

$$\langle A^k v, w \rangle = 0 \quad \forall k < \text{dist}(v, w)$$

Let  $(v, w)$  be a pair of vertices of maximal distance  $\delta$  apart.

$$\langle A^k v, w \rangle = \begin{cases} 0 & \text{if } k < \delta \\ \geq 1 & \text{if } k = \delta \end{cases}$$

Hence  $\text{span}(1, A, A^2, \dots, A^{\delta}, \dots)$  has

$$\text{dimension} \geq \delta + 1$$

On the other hand we have

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Lemma Let  $A$  be an endomorphism

which is diagonalisable, and let  $F[A]$

be the  $F$ -subalgebra of  $\text{End}(V)$

generated by  $A$ :  $F[A] = \text{span}(1, A, A^2, \dots)$

Then  $\dim_F F[A] \leq \# \text{spectrum}(A)$

Proof Let  $V = V_1 \oplus \dots \oplus V_\ell$  be the decomposition of  $V$  into  $A$ -eigenspaces with  $\ell = \# \text{spectrum}(A)$ . Then the map

$$\begin{aligned} F[A] &\longrightarrow F^\ell \\ T &\longmapsto (T|_{V_1}, T|_{V_2}, \dots, T|_{V_\ell}) \end{aligned}$$

is injective.



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## Regular graphs

The degree of a vertex is the number of vertices that are adjacent to it.

Definition: A graph is said to be regular if all its vertices have the same degree  $d$ .

$G$  is then called a  $d$ -regular graph.

Proposition If  $G$  is a connected  $d$ -regular graph, then  $d$  is the largest eigenvalue of  $A$ , and the eigenspace for  $d$  is one-dimensional, spanned by the constant functions on  $G$ .

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Proof Let  $f$  be an eigenfunction for  $A$ , and let  $v$  be the vertex where  $|f|$  attains its maximum

$$\begin{aligned} \text{Then } |(Af)(v)| &= |f(v_1) + \dots + f(v_d)| \\ &\leq |f(v_1)| + \dots + |f(v_d)| \leq d|f(v)|, \end{aligned} \quad (*)$$

where  $v_1, \dots, v_d$  are adjacent to  $v$ .

The last inequality uses the maximality assumption on  $v$ .

Hence  $|\lambda_1| \leq d$ . Furthermore, if  $Af = df$ , then we have equality in (\*) hence

$$\begin{aligned} |f(v_1)| + \dots + |f(v_d)| &= d|f(v)| \\ \Rightarrow |f(v_1)| = \dots = |f(v_d)| &= |f(v)| \end{aligned}$$

Furthermore,  $|f(v_1) + \dots + f(v_d)| = |f(v_1)| + \dots + |f(v_d)|$   
 $\Rightarrow f(v_1) = f(v_2) = \dots = f(v_d) = f(v)$ .

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It follows that any eigenfunction for  $A$  with eigenvalue  $d$  is proportional to the constant function  $1$ .  $\square$

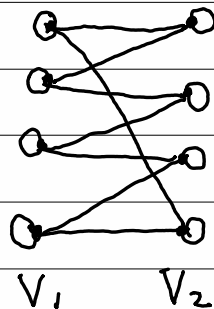
When is  $-d \in \text{Spec}(A)$ ?

Definition: A graph  $G$  is

bipartite if  $V(G) = V_1 \sqcup V_2$   
and

$$E(G) \subseteq V_1 \times V_2$$

Eg =



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Theorem: If  $G$  is bipartite,

then  $-d \in \text{Spectrum}(A)$

Proof: Let  $f(v) = \begin{cases} 1 & \text{if } v \in V_1 \\ -1 & \text{if } v \in V_2 \end{cases}$

clearly  $Af = -df$ .  $\square$

Fact: The converse is also true

(We will not prove this.)

The averaging operator

On a  $d$ -regular graph, it is

$$T = \frac{1}{d} A \quad Tf(v) = \frac{1}{d} \sum_{w \sim v} f(w)$$

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For example,  $V(G)$  could represent the members of a population, with edges describing the interactions between them.

A function  $f \in L^2(V(G))$  could represent the distribution of a certain resource (or virus!) being shared across the population.

$$T^k f \quad (k = 1, 2, 3, \dots)$$

is a simple model for how this distribution might evolve over time.



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Theorem: Let  $G$  be a connected  $d$ -regular graph which is not bipartite. If  $f \in L^2(V(G))$

$$\lim_{k \rightarrow \infty} T^k f(v) = \frac{1}{N} \left( \sum_{w \in V(G)} f(w) \right)$$

where  $N = \#V(G)$ .

Proof: Let  $\phi_1, \dots, \phi_N$  be

the eigenfunctions of  $T = \frac{1}{d} A$ ,

with eigenvalues  $1 = \lambda_1 > \lambda_2 \geq \dots \geq \lambda_N > -1$

By the spectral theorem, the

$\phi_j$  are pairwise orthogonal and

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can be chosen to be an orthonormal system for  $L^2(G)$ . Furthermore,

$$\phi_1(v) = \frac{1}{\sqrt{N}} \quad (\forall v \in V(G))$$

(i.e.,  $\phi_1$  is the appropriate multiple of the constant function, normalised so that  $\langle \phi_1, \phi_1 \rangle = 1$ .)

Now, write

$$f = \alpha_1 \phi_1 + \alpha_2 \phi_2 + \dots + \alpha_N \phi_N$$

with  $\alpha_1, \alpha_2, \dots, \alpha_N \in \mathbb{C}$ .

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$$T^k f = \alpha_1 \lambda_1^k \phi_1 + \alpha_2 \lambda_2^k \phi_2 + \dots + \alpha_N \lambda_N^k \phi_N$$

but  $\lambda_1 = 1$ , while  $|\lambda_2|, \dots, |\lambda_N| < 1$

$$\Rightarrow T^k f \longrightarrow \alpha_1 \phi_1 \quad \text{as } k \rightarrow \infty$$

$\alpha_1 \phi_1$  is the orthogonal projection of

$f$  onto  $\mathbb{C} \cdot \phi_1$ . Hence

$$\alpha_1 = \langle f, \phi_1 \rangle = \frac{1}{\sqrt{N}} \sum_{v \in V(G)} f(v)$$

$$\alpha_1 \phi_1 = \frac{1}{N} \left( \sum_{v \in V(G)} f(v) \right) \times \mathbb{1}_G$$

constant function  $\mathbb{1}$   
on  $V(G)$

$$\alpha_1 \phi_1(v) = \frac{1}{N} \sum_{v \in V(G)} f(v) \quad \square$$

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The size of the eigenvalues  
 $\lambda_2, \dots, \lambda_N$  controls the rate  
at which  $T^k f$  converges  
to the uniform distribution  
 $\alpha_1 \phi_1 \dots$

