

# Class Notes Starting Jan 2

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# CPSC 536J. Applied Linear Algebra

- Starting Monday, Jan 7, class 8:45-9:35 MWF
- Questionnaire on Friday regarding topics & times & grading
- Meaning of grades:

95%: Very Strongly to work in the area, promises that the instructor writes a strong letter of recommendation

90%: Strong ---

85%: { You need more basics to work in this field  
{ You learned a reasonable amount to use in another field

80%: You have fulfilled the expectations of a grad student

< 80%: " " not " " " " " " " "

Grades: Homework problems, collected { middle of term  
{ end of term

- Work on a project related to the course
- (1) pure research + expository article
  - (2) applied research, computations
  - (3) present to class some topic

First few weeks: what do the first few eigenvalues tell us and don't tell us about large systems?

- First topic:

- Symmetric matrices:  $A^T = A$ , more generally

- self-adjoint operators  $A^* = A$  on inner product spaces

- Abstract Theorem: Such matrices/operators have an orthonormal eigenbasis

- Examples:

- Expanding graphs

$[A = \text{Adj mat} : A^T = A]$

- Reversible Markov chains

- FA (factor analysis)  $\approx$  PCA (principle component analysis)

$\approx$  SVD (singular value decomposition)

- Variance-Covariance

- Laplacians of: graphs, simplicial complexes, etc.

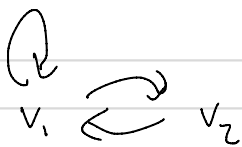
positive  
semi-  
definite,

$M^T M$

$=$

Example!

Fibonacci graph:



(directed graph, we allow self-loops, multiple edges)

$$\text{Adjacency matrix (Fib graph)} : \begin{matrix} & v_1 & v_2 \\ \begin{matrix} v_1 \\ v_2 \end{matrix} & \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \end{matrix} = A$$

$$(Adj)_{u,v} = \begin{cases} \# \text{ edges from } u \text{ to } v \end{cases}$$

$$A^2 = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}, A^3 = \begin{bmatrix} 3 & 2 \\ 2 & 1 \end{bmatrix}, A^4 = \begin{bmatrix} 5 & 3 \\ 3 & 2 \end{bmatrix}, \dots$$

$$F_1 = 1, F_2 = 1, F_3 = 2, \dots \quad F_n = F_{n-1} + F_{n-2}$$

$$A \begin{bmatrix} F_n \\ F_{n-1} \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} F_n \\ F_{n-1} \end{bmatrix} = \begin{bmatrix} F_n + F_{n-1} \\ F_n \end{bmatrix} = \begin{bmatrix} F_{n+1} \\ F_n \end{bmatrix}$$

Diagonalize: 2 eigenvalues  $\frac{1+\sqrt{5}}{2}$ ,  $\frac{1-\sqrt{5}}{2}$  ↙ between -1, 0

$$\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \frac{1+\sqrt{5}}{2} \\ 1 \end{bmatrix} = \left( \frac{1+\sqrt{5}}{2} \right) \begin{bmatrix} \phantom{\frac{1+\sqrt{5}}{2}} \\ \phantom{\frac{1+\sqrt{5}}{2}} \end{bmatrix}$$

(same)

We'll see  
 $F_n = \text{nearest integer to } \left( \frac{1+\sqrt{5}}{2} \right)^n \frac{1}{\sqrt{5}}$   
Quest:  $F_{-1}, F_{-2}, \dots$

Jan 4

- Today  $\rightarrow$  9:30 am

- Start next week 8:50 am  $\leftarrow$  MWF  
8:45 am

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Operators:  $T: V \rightarrow V$   $\left\{ \begin{array}{l} \leftarrow \text{inner product, dot product} \\ \leftarrow V = \mathbb{R}^n \end{array} \right.$   
 $\uparrow \quad \nearrow$   
 $\mathbb{R}^n$  for now

=

$T: \mathbb{R}^n \rightarrow \mathbb{R}^n$  think of it: standard basis  $e_1 = (1, 0, 0, \dots)$   
 $\vdots (0, 1, 0, \dots)$   
 $\vdots$   
 $e_n \quad \vdots$

$T \rightsquigarrow$  matrix  $\left[ \begin{array}{c} \overbrace{\quad}^n \\ \quad \end{array} \right] \}^n$

Abstractly:  $(\lambda, \vec{v})$  eigenpair for  $A \in \mathbb{R}^{n \times n}$  if  
eigen values  $\lambda$  vectors  $\vec{v}$

$$A\vec{v} = \lambda\vec{v}, \quad \vec{v} \neq 0.$$

If  $\lambda$  is eigenvalue,  $E_\lambda = \ker(A - \lambda I)^n$

generalized eigenspace of  $\lambda$

Thm: Each vector  $\vec{u} \in \mathbb{R}^n$  can be written uniquely as

Fix  $A \in \mathbb{R}^{n \times n}$ .

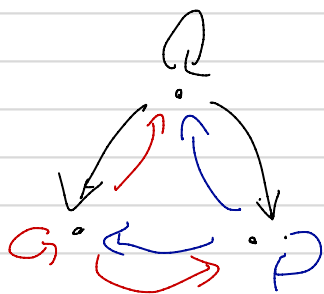
$$\vec{u} = \sum_{\lambda \in \text{Eigs}(A)} \vec{u}_\lambda, \quad \vec{u}_\lambda \in E_\lambda = E_\lambda(A)$$

Examples:

(0) Last time  $A = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$  adj mat of Fibonacci graph,

$$\lambda = \frac{1 \pm \sqrt{5}}{2}$$

(1) Complete digraph on 3 vertices



$$A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

$$\lambda's: A \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$\lambda's: 0's$

because: C. :  $A = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}$  rank 1

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \end{bmatrix} + 0$$

$$E_0 = \left\{ \begin{bmatrix} a \\ b \\ c \end{bmatrix} \mid a+b+c=0 \right\} = \text{Vects } \perp \text{ to } \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \text{ i.e. } \begin{bmatrix} a \\ b \\ c \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = 0$$

$$A \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} a+b+c \\ a+b+c \\ a+b+c \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \quad \left\{ v \mid A\vec{v} = 0 \cdot \vec{v} \right\}$$

Last time: Any symmetric  $A$  ( $A^T = A$ ) has  
 orthonormal eigenbasis  $\vec{v}_1, \dots, \vec{v}_n$  s.t.  $A\vec{v}_i = \lambda_i \vec{v}_i$

and  $\vec{v}_i \cdot \vec{v}_j = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{otherwise} \end{cases}$

then

$$A = \sum_{i=1}^n \lambda_i \underbrace{\begin{pmatrix} \vec{v}_i & \vec{v}_i^T \end{pmatrix}}_{\text{rank 1}}$$

=  
 e.g.  $A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$ ,  $\lambda_1 = 3$ ,  $\vec{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} / \sqrt{3} = \begin{bmatrix} 1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{bmatrix}$

$\lambda_2, \lambda_3 = 0$ ,  $\vec{v}_2 = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \\ 0 \end{bmatrix}$   $\vec{v}_3 = \begin{bmatrix} 1/\sqrt{6} \\ 1/\sqrt{6} \\ -2/\sqrt{6} \end{bmatrix}$

(☹)                      (☹)

$$A = 3 \cdot \vec{v}_1 \vec{v}_1^T + 0 \cdot \vec{v}_2 \vec{v}_2^T + 0 \cdot \vec{v}_3 \vec{v}_3^T$$

=  
 $\begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix} = A - I = \text{adj}_\Delta = 2 \vec{v}_1 \vec{v}_1^T + (-1) \vec{v}_2 \vec{v}_2^T + (-1) \vec{v}_3 \vec{v}_3^T$

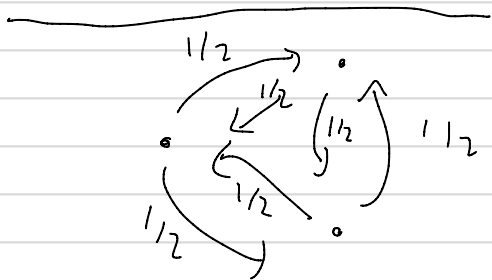
Question: Is there a nice way to write

$$\begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix} = (2) \begin{bmatrix} ? \\ ? \\ ? \end{bmatrix} + (-1) \begin{bmatrix} ? \\ ? \\ ? \end{bmatrix}$$

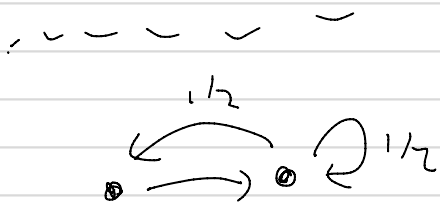
Proj onto orthog. comp  
of  $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$

$$= E_{-1} \left( \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix} \right)$$

$$= E_0 \left( \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \right)$$



$$\begin{bmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{bmatrix}$$



$$\begin{bmatrix} 0 & 1 \\ 1/2 & 1/2 \end{bmatrix}$$



Why do we write  
 $A^* = A$   
vs.  $A^T = A$



Jan 7:

If  $A \in \mathbb{R}^{n \times n}$ , i.e.  $A$  is  $n \times n$  real matrix:

eigenpair for  $A$ :  $(\lambda, \vec{v})$  s.t.  $A\vec{v} = \lambda\vec{v}$   
eigenvalue ← eigenvalue  
eigenvector ← eigenvector

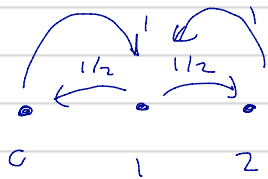
First few weeks: think of  $n \times n$ ,  $n$  large

$A \leftrightarrow$  local structure: if eigenvalues of  $A$ :  $\lambda_1, \dots, \lambda_n$

have, say  $\lambda_1, \lambda_2, \lambda_3$  big,  $\lambda_4, \dots, \lambda_n$  small(er), can we say useful things about  $A$ ?

Examples: (1) Adjacency matrices of graphs / digraphs  
symmetric matrices

(2) Markov chains:



$$P_{0 \rightarrow 1} = 1 = P_{2 \rightarrow 1}$$

$$P_{1 \rightarrow 2}, P_{1 \rightarrow 0} = 1/2$$

Markov matrix is one with stochastic rows:

$$P = \begin{pmatrix} P_{00} & P_{01} & P_{02} \\ P_{10} & P_{11} & P_{12} \\ P_{20} & P_{21} & P_{22} \end{pmatrix}$$

$$P_{21} = P_{2 \rightarrow 1}$$

$$P = \begin{pmatrix} 0 & 1 & 0 \\ 1/2 & 0 & 1/2 \\ 0 & 1 & 0 \end{pmatrix}$$

Markov chains: (1) What happens to  $P^1, P^2, \dots, P^n$ ,  $n \rightarrow \infty$

(2) For what  $n$  is  $P^n$  close to  $P^\infty := \lim_{n \rightarrow \infty} P^n$ ?

(3) Often matrices  $A \in \mathbb{R}^{n \times n}$  with  $A^T = A$  (symmetric)  
(SVD, FA, PCA, Laplacians, ...)

General Theory:  $A \in \mathbb{R}^{n \times n}$  identity matrix  
↓

$$A\vec{v} = \lambda\vec{v}, \quad \vec{v} \neq \vec{0} \quad \Leftrightarrow \quad A\vec{v} = \lambda I\vec{v}$$

$$\Leftrightarrow (A - \lambda I)\vec{v} = \vec{0}$$

$\lambda$  eigenvalue of  $A \Leftrightarrow A - \lambda I$   $\begin{cases} - \text{is not invertible} \\ - \text{has non-zero kernel/nullspace} \\ - \text{has zero determinant} \\ \vdots \end{cases}$

$$\Leftrightarrow \det(xI - A) = \underset{\text{poly}}{\text{char}}(A)(x) \quad \text{has root } \lambda.$$

Thm:  $\det(xI - A) = p_A(x)$  has  $n$  roots, and if

$\lambda$  is a root of multiplicity  $k$ , then  $E_\lambda(A) := \ker(\lambda I - A)^n$

has dim  $k$ .

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$$\text{Eg. } A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \quad \det(xI - A) = \det \begin{bmatrix} x & -1 \\ 0 & x \end{bmatrix} = x^2$$

$$\lambda\text{'s of } A : 0, 0. \quad \lambda I - A \Big|_{\lambda=0} = -A = \begin{bmatrix} 0 & -1 \\ 0 & 0 \end{bmatrix} \quad \text{☹}$$

$$\left\{ \vec{v} \mid A\vec{v} = 0 \cdot \vec{v} \right\} = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$$

$$\left\{ \vec{v} \mid (A - I \cdot 0)^{\text{pos power}} \vec{v} = \vec{0} \right\} : \quad A^2 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

"defect of  $\lambda=0$ "  $\Leftrightarrow \ker(I\lambda - A) = \text{Eigenvectors}$   
is 1-dimensional  $\lambda \neq 0$

Fig.  $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ ,  $B = A + 3I = \begin{bmatrix} 3 & 1 \\ 0 & 3 \end{bmatrix}$

$$A\vec{v} = \lambda\vec{v} \iff B\vec{v} = (\lambda+3)\vec{v}$$

matrix  $\begin{bmatrix} \lambda & 1 & & 0's \\ & \lambda & \ddots & \\ & & \ddots & 1 \\ 0's & & & \lambda \end{bmatrix}$  "Jordan block"

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Any matrix is similar to a block diagonal matrix whose blocks are Jordan blocks

$$\begin{array}{l} 2 \{ \\ 2 \{ \\ 1 \{ \end{array} \left[ \begin{array}{c|c|c} \begin{matrix} 3 & 1 \\ 0 & 3 \end{matrix} & 0's & 0's \\ \hline 0 & \begin{matrix} 5 & 1 \\ 0 & 5 \end{matrix} & 0 \\ \hline 0 & 0 & 2 \end{array} \right]$$

$\underbrace{\hspace{2em}}_2 \quad \underbrace{\hspace{2em}}_2 \quad \underbrace{\hspace{1em}}_1$

$A, B$   $n \times n$  are similar if  $B = M^{-1} A M$

$(A \sim B)$

for  $M$  invertible

①  $A \sim B \iff A, B$  differ by a change of basis

$B = M^{-1} A M$

$$\begin{aligned} \implies B^{100} &= \underbrace{M^{-1} A M}_{\text{we understand } A} \underbrace{M^{-1} A M}_{\text{we understand } A} \dots M^{-1} A M \\ &= M^{-1} A^{100} M \end{aligned}$$

(we understand  $A$ )

$$B = M^{-1} A M \quad (\mathbb{R}^{n \times n})$$

$$p = \text{polynomial} \quad p(x) = 2x^2 + x^4$$

$$p(B) = 2B^2 + B^4$$

$$2(M^{-1} A M)^2 + (M^{-1} A M)^4 = M^{-1} (2A^2 + A^4) M$$

$$p(B) = M^{-1} p(A) M \quad \leftarrow \text{for any reasonable } p$$

$$\text{If } A = \begin{bmatrix} d_1 & & 0 \\ & d_2 & \\ 0 & & \ddots \\ & & & d_n \end{bmatrix}, \quad A^2 = \begin{bmatrix} d_1^2 & & \\ & d_2^2 & \\ & & \ddots \\ & & & d_n^2 \end{bmatrix}$$

$$p(A) = \begin{bmatrix} p(d_1) & & 0's \\ & \ddots & \\ 0's & & p(d_n) \end{bmatrix}$$

$$\text{Def: } B \text{ is diagonalizable iff } B = M^{-1} \begin{bmatrix} d_1 & & 0's \\ & \ddots & \\ 0's & & d_n \end{bmatrix} M$$

$$\text{Then } p(B) = M^{-1} \begin{bmatrix} p(d_1) & & 0's \\ & \ddots & \\ 0's & & p(d_n) \end{bmatrix} M$$

$\leftarrow$  We'll see:  
 if  $B^T = B$ ,  $M$  is  
orthogonal matrix, things  
 work nicely; otherwise  
 $M, M^{-1}$  "not nice"

Examples:

Fibonacci graph



adjacency  
mat

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

Symm

$$A = M^{-1} \begin{pmatrix} d_1 & 0 \\ 0 & d_2 \end{pmatrix} M$$

$$d_1, d_2 = \frac{1 \pm \sqrt{5}}{2}$$

$$A^n, n \text{ large} \approx M^{-1} \begin{pmatrix} d_1^n & 0 \\ 0 & \text{close to } 0 \end{pmatrix} M$$

$$A^n, n \text{ very negative} \approx M^{-1} \begin{pmatrix} \text{close to } 0 & 0 \\ 0 & d_2^n \end{pmatrix} M$$

eigenvectors:

$$\begin{pmatrix} \frac{1+\sqrt{5}}{2} \\ 1 \end{pmatrix}, \begin{pmatrix} \frac{1-\sqrt{5}}{2} \\ 1 \end{pmatrix}$$

$$\Leftrightarrow r^2 = r + 1$$

$$\Leftrightarrow F_n = F_{n-1} + F_{n-2}$$

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Recall: If  $A^T = A$ , then  $A\vec{v} = \lambda\vec{v}$ ,  $\lambda \neq \mu$ ,  $\vec{v} \cdot \vec{u} = 0$   
 $A\vec{u} = \mu\vec{u}$

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

$$\lambda = 3, 0, 0$$

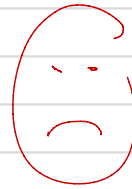
$$\sum_{\lambda \in \text{eigs}(A)} \lambda \cdot (\text{proj onto eigenspace of } \lambda)$$



$$\lambda_1 = 3, \quad \vec{v}_1 = \begin{bmatrix} 1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{bmatrix}$$

$$\lambda_2 = 0, \quad \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \\ 0 \end{bmatrix}$$

$$\lambda_3 = 0, \quad \begin{bmatrix} 1/\sqrt{8} \\ 1/\sqrt{8} \\ -2/\sqrt{8} \end{bmatrix}$$



$\mathbb{R}^3$

orthonormal

$$\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$$

$$A \vec{v}_i = \lambda_i \vec{v}_i$$

$$\sum_i \lambda_i \left( \vec{v}_i \vec{v}_i^T \right)$$

Next time:

- Markov matrices

- Matrices like

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

$$\lambda = 3, 0, 0$$

and matrices of  $d$ -regular graphs

expansion bounds

connected?

connectivity?

Jan 9: Today: Edge expansion, Markov chains

Questions (courtesy of Carl (et al.?):

(1)  $\sin(A) = A - \frac{A^3}{3!} + \frac{A^5}{5!} - \dots$  What norm? (Short answer: doesn't matter)

(2) If  $A \neq 0$  is nilpotent, is it not diagonalizable? (Short answer: yes)

=  
If  $A \in \mathbb{R}^{m \times n}$ , typical norms:

(1)  $\|A\|_{\text{Frob}} = \sqrt{\text{Tr}(A^T A)} = \sqrt{\sum_{i,j} (A_{ij})^2}$  entries of  $A$

$\left[ \begin{array}{l} \text{Tr}(M), M \text{ square, } \text{Tr}(M) = \sum_i M_{ii} = \sum_{i=1}^n \lambda_i, \dots \\ \text{Tr}(A^T B) = \text{Tr}(B^T A), \dots \end{array} \right]$

Operator norm:  $L: V \rightarrow V$ , same norm on  $V$ ,

Norm:  $\|\cdot\|: V \rightarrow \mathbb{R}_{\geq 0}$  ( $\mathbb{R}$ -vector sp)  
( $\mathbb{C}$ - " " )

(1)  $\|\alpha v\| = |\alpha| \|v\|$

(2)  $\|\cdot\|$  is a metric: (1)  $\|v\| \geq 0$  and eq iff  $v=0$

(2)  $\|v+w\| \leq \|v\| + \|w\|$

$\rho_{\|\cdot\|}(v,u) := \|v-u\|$

(2)  $\|L\| := \max_{v \neq 0} \frac{\|Lv\|}{\|v\|}$   
wrt some  $\|\cdot\|$  on  $V$



Remark: If  $V$  is a finite dim  $\mathbb{R}$ -vector space ( $\mathbb{C}$ -vector space)

Then for any norms  $\|\cdot\|_1, \|\cdot\|_2$ , there are  $C', C$  st.

for all  $v \in V$

$$0 < C' \leq \frac{\|v\|_{\text{first}}}{\|v\|_{\text{second}}} \leq C$$

( $C', C$  depend on  $\|\cdot\|_1, \|\cdot\|_2$ )

So  $v_1, v_2, v_3, \dots \in V$ :

$$\lim_{i \rightarrow \infty} \|v_i\|_{\text{first}} = 0 \iff \lim_{i \rightarrow \infty} \|v_i\|_{\text{second}} = 0$$

Proof: Take a basis for  $V \iff \mathbb{R}^n$

$v_1, \dots, v_n$

$\Rightarrow$  each elt of  $V$  is

uniquely rep as

$$\alpha_1 v_1 + \dots + \alpha_n v_n$$

$$\rightarrow \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix}$$

$$\begin{aligned} & |\alpha_1|^3 + |\alpha_2|^3 + \dots + |\alpha_n|^3 \\ & = 27.215 \end{aligned}$$

supremum / infimum of  $\frac{\|v\|_{\text{first}}}{\|v\|_{\text{second}}}$  is same as max/min on  $\left\{ v \sim \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{bmatrix} \mid \alpha_1^2 + \dots + \alpha_n^2 = 1 \right\}$

continuous in  $V$

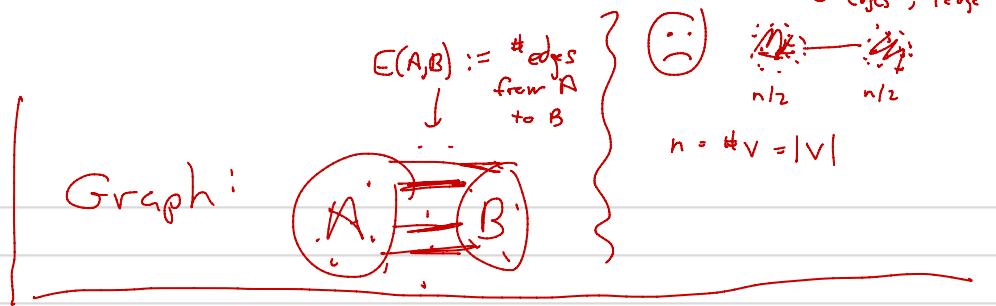
compact, non-zero vectors

infimum (greatest lower bound) is attained at some vector  $u \in V$

$$\text{Hence } \vec{u} \neq 0, \quad 0 \neq \frac{\|u\|_1}{\|u\|_2} \leq \text{all } \frac{\|v\|_1}{\|v\|_2}$$



Edge expansion:



Thm: Let  $G$  be a  $d$ -regular graph on  $n$  vertices.

(so  $d = \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n \geq -d$ ). If  $A, B \subset \bar{V}$ , then

$$\left| \left( \# \text{ edges from } A \text{ to } B \right) - \frac{d}{n} |A| |B| \right|$$

not tight }  $\rightarrow \leq \rho \sqrt{\frac{|A|(n-|A|)}{n}} \sqrt{\frac{|B|(n-|B|)}{n}}$

OK for  $|A|, |B|$  small }  $\rightarrow \leq \rho \sqrt{|A| \cdot |B|}$  where  $\rho = \max_{i>1} |\lambda_i|$

but  $|A| \cdot |B|$  not too small

Jan 11:

Graph  $G = (V, E)$ :

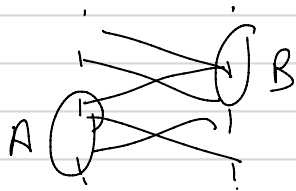
$$A, B \subset \bar{V}, \quad (1) E(A, B) = \left\{ \begin{array}{l} \text{set} \\ \text{edges joining} \\ A \text{ \& B} \end{array} \right.$$

$$(2) e(A, B) = |E(A, B)|$$



Bipartite graph

$$(3) \Gamma(A) := \left\{ v \in V \mid \begin{array}{l} v \notin A \\ v \text{ has edge to} \\ \text{an element of } A \end{array} \right.$$



In classical notion of expander

(for switching networks)

- given  $A$ , want  $|\Gamma(A)|$  large

- given  $A, B$  want  $e(A, B)$  large

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Fix  $d$ , think of  $n \rightarrow \infty$ , want

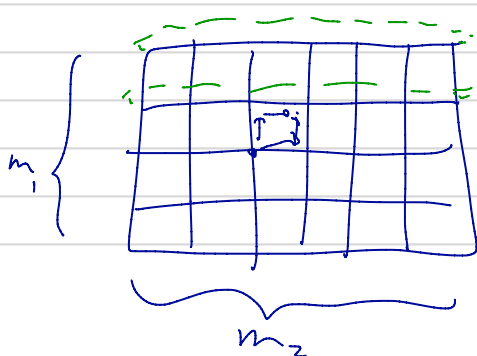
-  $d$ -regular graph

-  $d$ -regular bipartite graph

with good  $\left\{ \begin{array}{l} \text{edge expansion} \\ \text{vertex expansion} \end{array} \right.$

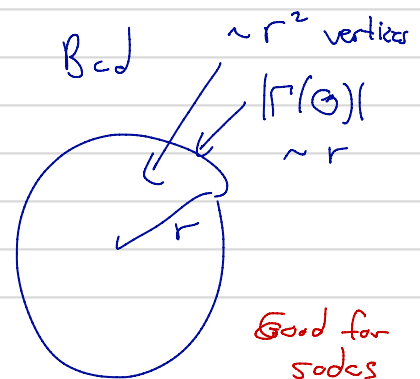
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Bad expander:  $d=4$



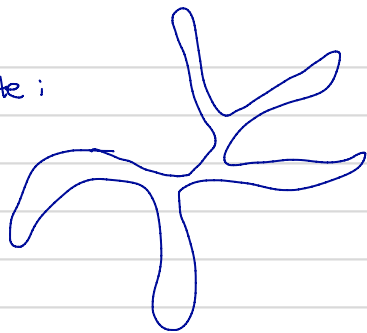
Grid  $\mathbb{Z}/m_1\mathbb{Z} \times \mathbb{Z}/m_2\mathbb{Z}$

+ wrap around



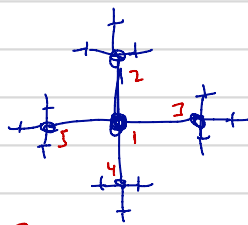
Good for nodes

Opposite:



no hotspot  
in going inside  
to outside

Say  $d=4$



best  
locally expander

locally tree

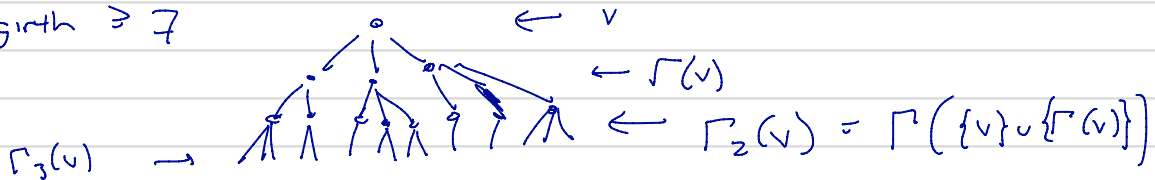
at  
 $\sim O(\sqrt{n})$   
labels, see  
repeat

$$V = \{1, \dots, n\}$$

Building a graph that is a local tree is easy.

$\text{Girth}(\text{graph}) := \text{length of shortest cycle}$

if  $\text{girth} \geq 7$

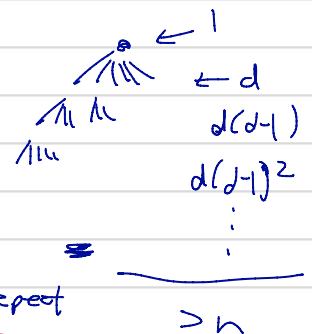


Open Problem: Given fixed  $d$ , large  $n$

$$\text{Girth}(\text{any graph } d\text{-reg on } n\text{-vertices}) \leq (2+o(1)) \log_{d-1}(n)$$

Moser bound 1950's

precise  
value



repeat

> n

Random graph

with corrections  
of a few bad edges

$$\text{girth} = (1+o(1)) \log_{d-1}(n)$$

LPS-Margulis (bipartite)  
expanders

$$\frac{4}{3} \log_{d-1}(n)$$

$d = 1 + \text{prime}$

Open: Fix any integer  $d \geq 3$ . Is

$$\text{growth} \left( \begin{array}{l} \text{Add } d\text{-reg graphs} \\ \text{or } n \text{ vertices} \end{array} \right) \leq (2 - \varepsilon) \log_{d-1}(n)$$

for some  $\varepsilon > 0$ ,  $n$  suff large.

==

**Vertex Expander:**  $(\alpha, \beta, \gamma, \delta, \varepsilon)$ -expander:  $G$ ,  $d$ -regular on  $n$  vert

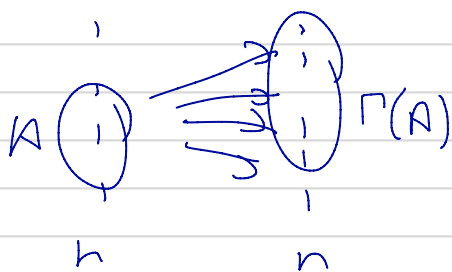
$$\forall A \subset V : \text{if } |A| \leq \alpha n$$

$$|\Gamma(A)| \geq \beta |A|$$

$$\text{if } n\gamma \leq |A| \leq n\delta$$

$$\text{then } |\Gamma(A)| \geq \varepsilon |A|$$

Vertex Expansion:  $d$ -reg (bipartite) graph on  $n$ -vertices s.t.



$$|\Gamma(A)| \geq 2|A|$$

for all  $A \subset V$  such that

$$|A| \leq \frac{n}{3}$$

You have  $d$  fixed,  $n \rightarrow \infty$ , choose  $d$  random permutations

$$\text{on } [n] = \{1, \dots, n\}$$

- Edge expansion

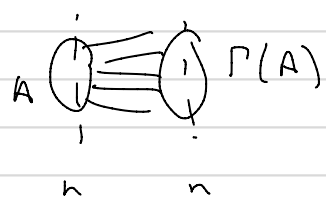
- Eigenvalue expansion ←

Jan 14:

Last time: 2 open problems:

1950's { ① Moore bound:  $\text{girth} \left( \begin{smallmatrix} \text{any } d\text{-reg} \\ \text{graph on } n \\ \text{vertices} \end{smallmatrix} \right) \leq 2 \log_{d-1} n + o(1)$   
Does  $2 \rightarrow 2-\epsilon$  hold?

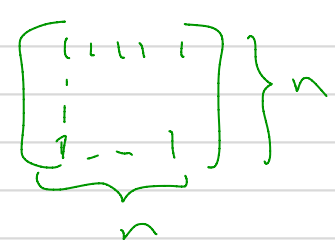
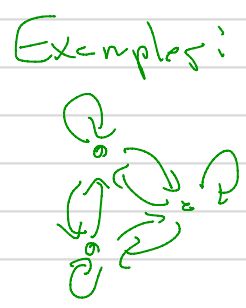
more is known { ② Construct bipartite graph  $d$ -regular with smallest  $d$ ,  $n \rightarrow \infty$   
s.t. for all  $A \subset$  left vertices,  $|A| \leq \frac{n}{3}$   
 $|\Gamma(A)| \geq 2|A|$ .



Both fundamentally connected to eigenvalues of Adjacency(G).

Today: ① Examples of Graphs & Eigenvalues  
② More on "Expansion"

Classical Examples of Good/Bad Connectivity in Graphs:

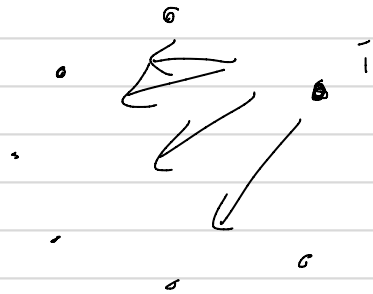


$\lambda_1 = n$  mult. 1  
 $0$  mult.  $n-1$   
 $\lambda_1 = n \geq \lambda_2 \geq \dots \geq \lambda_n$

😊 Connected

☹️ degree  $d=n$

② Graph:  $V = \mathbb{Z}/p\mathbb{Z} = \{0, \dots, p-1\}$



For  $i \in V$ , edge

$i \rightarrow i+j$  where  $j$  is a quadratic residue,

Def: We say that  $j$  is a quadratic residue mod  $p$

if  $j \not\equiv 0 \pmod{p}$

and  $j \equiv a^2 \pmod{p}$  for some  $a$ .  $a^2 = (-a)^2$ ,

=

e.g.  $p=11$ ,  $1^2 \equiv 1 \pmod{11}$ ,  $2^2 \equiv 4 \pmod{11}$ ,  $3^2 \equiv 9 \pmod{11}$ ,

$4^2 \equiv 5 \pmod{11}$ ,  $5^2 \equiv 3 \pmod{11}$

$6^2 \equiv (-5)^2$  done,  $7^2 \equiv (-4)^2$  done, ...

Quadratic residues mod 11 : 1, 3, 4, 5, 9 mod 11

" non-residues " " : 2, 6, 7, 8, 10 mod 11

Fact: res  $\cdot$  res  $\equiv$  res

(non-res)  $\cdot$  res  $\equiv$  non-res

(non-res) (non-res)  $\equiv$  res

$2^0, 2^1, 2^2, \dots, 2^{10}$  get all non-zero mod 11  
1, 2, 4, 8, 5, ... 6



Homework: ① If  $p$  is prime,  $i \neq 0$ ,  $x^2 \equiv i \pmod{p}$  has either 2 solutions or no solutions

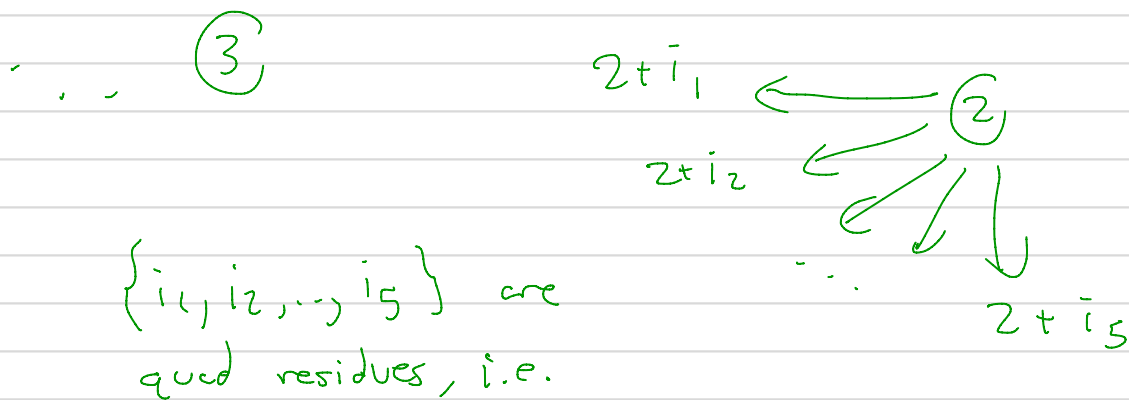
② There are  $\frac{p-1}{2}$  values of  $i$  s.t.  $x^2 \equiv i \pmod{p}$  has 2 solutions [such  $i$  we called Quadratic Residues]

③

res  $\cdot$  res  $\equiv$  res  
 (non-res)  $\cdot$  res  $\equiv$  non-res  
 (non-res) (non-res)  $\equiv$  res

=

⑩ ⑨ ①  $p=11$   
 $\leftarrow V = \mathbb{Z}/p\mathbb{Z} = \{0, 1, \dots, p-1\}$



$\{1, 3, 4, 5, 9\}$

Claim: This graph is  $d$ -regular, with  $d = \frac{p-1}{2}$  (here 5)

$$d = \lambda_1 \geq \underbrace{\lambda_2 \geq \dots \geq \lambda_p}_{\text{small}}$$

😊 good "expansion"

☹  $d = \frac{p-1}{2} \approx \frac{n}{2}$

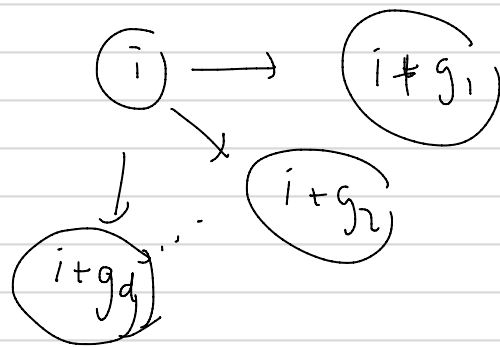
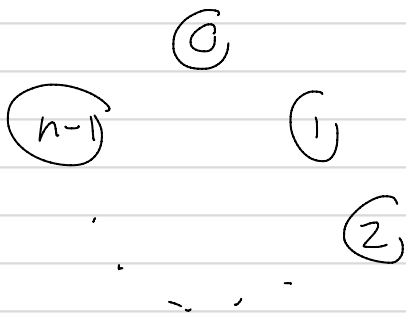
Fact! Say:

$$\text{Graph} = \text{Cayley}(\mathbb{Z}/n\mathbb{Z}, \{g_1, \dots, g_d\})$$

"generators"  
 $\downarrow \downarrow \downarrow$

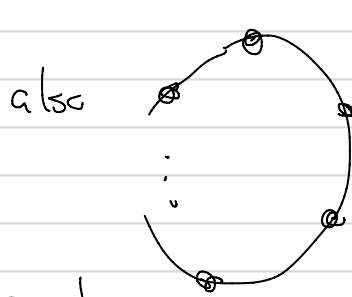
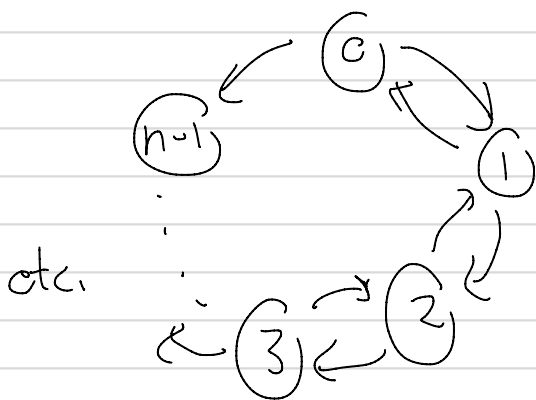
Vertices:  $\mathbb{Z}/n\mathbb{Z}$

$g_1, \dots, g_d \in \mathbb{Z}/n\mathbb{Z}$



Example:

$$\text{Cayley}(\mathbb{Z}/n\mathbb{Z}, \{1, -1\})$$



bad  
 expander

cycle  
 length  
 $n$

If  $\zeta^n = 1$ , i.e.  $\zeta$  is an  $n$ -root of unity, i.e.  $\zeta = \left( e^{\frac{2\pi i}{n}} \right)^k$   
 Then  $f: V \rightarrow \mathbb{C}$ ,  $f(j) = \zeta^j$  is an eigenfunction

$$\zeta^n = 1 :$$

$$\zeta^0 = \zeta^n = 1$$

$$\zeta^{n-1} \quad (n-1)$$

$$(1) \quad \zeta^1$$

$$(2) \quad \zeta^2$$

$$\zeta^{j+g_1}$$

$$(j+g_1)$$

$$\zeta^j$$

$$(j)$$

$$(j+g_2)$$

$$\zeta^{j+g_2}$$

$$(j+g_d)$$

$$\zeta^{j+g_d}$$

$$(Adj \text{ op on } f)(j)$$

$$= \sum_{j' \sim j} f(j')$$

$$\text{if } f = \begin{bmatrix} \zeta^0 \\ \zeta^1 \\ \vdots \\ \zeta^{n-1} \end{bmatrix} = \zeta^j \left( \underbrace{\zeta^{g_1} + \zeta^{g_2} + \dots + \zeta^{g_d}} \right)$$

$$S_G \quad Adj \begin{bmatrix} \zeta^0 \\ \zeta^1 \\ \vdots \end{bmatrix} = \left( \quad \right) \begin{bmatrix} \zeta^0 \\ \zeta^1 \\ \vdots \end{bmatrix}$$

Jan 16:

Multiply:  $p(x) = a_0 + a_1x + a_2x^2 + \dots + a_nx^n$

$q(x) = b_0 + \dots + b_nx^n$

Naïve way to compute  $p(x)q(x)$  is  $O(n^2)$  } FLOPs  
Time  
:

Imagine you can compute  $p(x), q(x)$  at  $2^m$ -th roots of unity  $z^n \Rightarrow 2n$  (or any # parts  $\geq 2n$ )

Can find  $p(\gamma_1), \dots, p(\gamma_{2n})$  "quickly"  
 $q(\gamma_1), \dots, q(\gamma_{2n})$

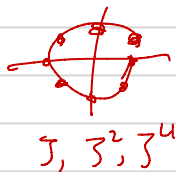
can compute

$(pq)(\gamma_1), \dots, (pq)(\gamma_{2n})$  in  $2n$  mults

Say given deg  $2n$  poly  $r = r(x)$ , and given  $r(\gamma_1), \dots, r(\gamma_{2n})$  you can find  $r(x)$  quickly

"Fast discrete Fourier transform" for certain  $\gamma_1 \dots \gamma_{2n}$  roots of unity,

Image:  $\zeta^8 = 1$



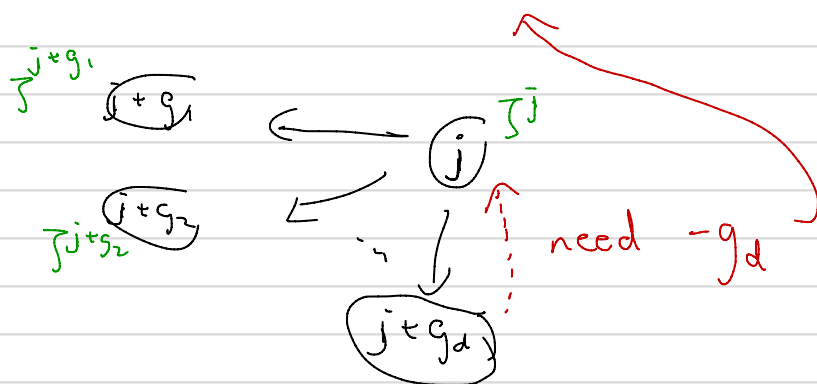
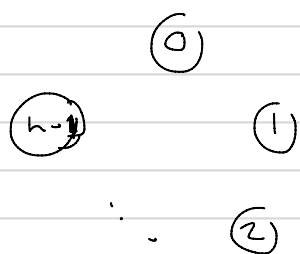
$\zeta = -\zeta^5$

$\zeta^m = 1$   
Some  $m$   
 $e^{2\pi i x}$

For us:

$$\mathbb{Z}/h\mathbb{Z}$$

$$G = \text{Cayley}(\mathbb{Z}/h\mathbb{Z}, \{g_1, \dots, g_d\}) :$$

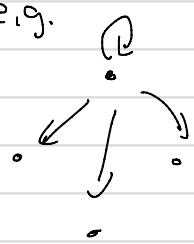


Claim: for any  $\vec{\lambda}$  s.t.  $\vec{\lambda}^n = 1$ , eigenvector

$$\text{Adj}_G \begin{bmatrix} \lambda^0 \\ \lambda^1 \\ \vdots \\ \lambda^{h-1} \end{bmatrix} = (\lambda^{g_1} + \dots + \lambda^{g_d}) \begin{bmatrix} \lambda^0 \\ \vdots \\ \lambda^{h-1} \end{bmatrix}$$

=

e.g.

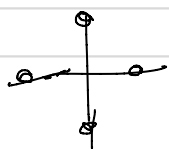


complete digraph on  $0, 1, 2, 3$  (4 vertices)

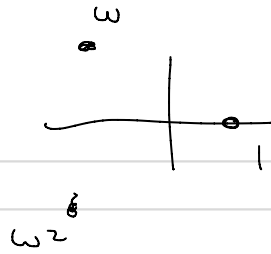
$$\text{Adj}_G = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$

$\lambda = 4, 0, 0, 0$   
rank 1

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} i^0 \\ i^1 \\ i^2 \\ i^3 \end{bmatrix} = \begin{bmatrix} \text{all } 1\text{'s} \end{bmatrix} \begin{bmatrix} 1 \\ i \\ -1 \\ -i \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} = 0 \begin{bmatrix} 1 \\ i \\ -1 \\ -i \end{bmatrix}$$



Similarly  $\lambda = 3, 0, 0$   
 $\omega = e^{2\pi i/3}$



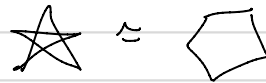
$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & \omega & \omega^2 \\ 1 & \omega^2 & \omega \end{bmatrix} \begin{bmatrix} 1 \\ \omega \\ \omega^2 \end{bmatrix} = 0$$

$$\text{" } \begin{bmatrix} 1 \\ \omega^2 \\ \omega \end{bmatrix} = 0$$

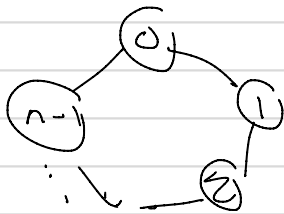
$$\text{" } \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \\ 3 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

Cayley( $\mathbb{Z}/n\mathbb{Z}$ ,  $\mathbb{Z}/n\mathbb{Z}$   
 "  $\{0, 1, \dots, n-1\}$ )

Worst Connected Graph Regular  
 in term expansion:



Cayley( $\mathbb{Z}/n\mathbb{Z}$ ,  $\{1, -1\}$ )



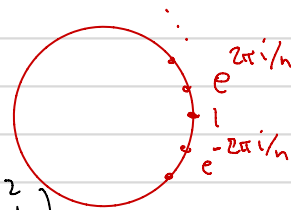
cycle  
of length  
 $n$

Formula says: take any  $\zeta$   
 with  $\zeta^n = 1$ ,  $\lambda = \lambda_\zeta = \zeta + \zeta^{-1}$

$\lambda_1 =$  largest eigenvalue  $= 2$ ,

$\lambda_2 =$  next largest  $=$   
 $e^{2\pi i/n} + e^{-2\pi i/n}$

$$= 2 \cos\left(\frac{2\pi}{n}\right) = 2 \left(1 - \left(\frac{2\pi}{n}\right)^2 + \dots\right) = 2 - \text{order}\left(\frac{1}{n^2}\right)$$

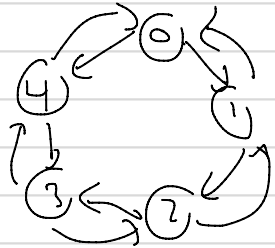


$p$  prime Cayley  $(\mathbb{Z}/p\mathbb{Z}, \{ \text{all quadratic residues} \})$

OR

" " " " non-residues

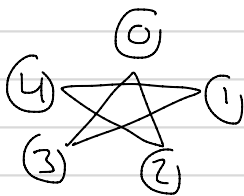
$p=5$



$$1^2 \equiv 1, 2^2 \equiv 4, 3^2 \equiv 4, 4^2 \equiv 1 \pmod{5}$$

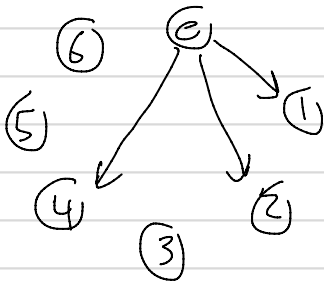
quad residues  $1, -1$

non-residues  $2, -2$



$p=7$

$$1^2 \equiv 1, 2^2 \equiv 4, 3^2 \equiv 2, 4^2 \equiv 2, 5^2 \equiv 4, 6^2 \equiv 1$$



$-1$  is not a quad residue

Fact:  $-1$  is quad residue mod  $p$ ,  $p$  odd prime iff  $p \equiv 1 \pmod{4}$

Degree:  $\frac{p-1}{2}$

Question: If  $\sum^p = 1$ , what is  $\sum_{\substack{r \in \text{quad} \\ \text{res mod } p}} \int^r$  for  $p \equiv 1 \pmod{4}$

Homework:  $p$  prime (maybe  $p \equiv 1 \pmod{4}$ )

What is  $\lambda_{\zeta} = \frac{1}{2} \sum_{\substack{a \neq 0 \\ (\text{mod } p)}} \zeta^{a^2}$ , for  $\zeta \neq 1$  ?

$$\lambda_1 = \frac{p-1}{2}. \quad \zeta \neq 1 \text{ but } \zeta^p = 1$$

Hint

$$\begin{aligned} \left( \sum_{a \neq 0} \zeta^{a^2} \right)^2 &= \sum_{a \neq 0} \zeta^{a^2} \sum_{b \neq 0} \zeta^{b^2} \\ &= \sum \zeta^{a^2 - b^2} \\ &= \sum_{a, b \neq 0} \zeta^{(a+b)(a-b)} \\ &= \sum_{\{\alpha, \beta\}} \zeta^{\alpha \beta} \end{aligned}$$



Jan 18:

$$\begin{bmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{bmatrix}$$

d-regular graph,  
n vertices:

d = n <sup>big</sup> 😞

row sums  
= col sums

$$\lambda_1 = n = d \geq \lambda_2 \geq \dots \geq \lambda_n \geq -d$$

here  $\lambda_2, \dots, \lambda_n = 0$  😊

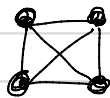
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Variant

$$\begin{bmatrix} 0 & 1 & \dots & 1 \\ & 0 & & 1 \\ & 1 & 0 & 0 \\ & & & \ddots \end{bmatrix}$$

complete  
graph

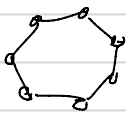
$$= A_{\text{complete digraph}} - I$$



$$\lambda_1 = d = n-1, \lambda_2 = \dots = \lambda_n = -1$$

=

Cycle



d=2,  $\lambda$ 's:  $2 \left( \cos \frac{2\pi}{n} \cdot m \right)$   $m=0, \dots, n-1$

$$\lambda_1 = d = 2 \geq \lambda_2 = 2 \cos \frac{2\pi}{n} = 2 - \text{order} \left( \frac{1}{n^2} \right)$$

$$\frac{2 \cdot (2\pi)^2}{2!} \frac{1}{n^2} + O\left(\frac{1}{n^4}\right)$$

=

Given d-regular graph on n vertices, let

$$\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$$

$\lambda_i = \lambda_i(G)$  are the eigenvalues of  $A =$  Adjacency matrix of  $G$  graph on  $n$  vertices. ~~Fix integer~~  $d \geq 3$ . Any graph on  $n$

=

$$\lambda_2 \geq 2\sqrt{d-1} \left( 1 - \left( \frac{1}{\log_{d-1}(n)} \right)^2 \right)$$

Fix  $d \geq 3$ . For a graph  $G$   $d$ -regular on  $n$  vertices, we

have

$$\lambda_2(G) \geq 2\sqrt{d-1} \left( 1 - O\left(\frac{1}{\log_{d-1} n}\right)^2 \right)$$

i.e.

$$2\sqrt{d-1} (1 + o(1))$$

[log<sub>d-1</sub> familiar from  
graph/Moore bound]

Thm: Fix  $d \geq 3$ . Then for any  $\epsilon > 0$ ,

for "most"  $d$ -regular graphs on  $n$  vertices, as  $n \rightarrow \infty$

with high probability for a random graph - - - - -

WHP

$$\lambda_2 \leq 2\sqrt{d-1} + \epsilon$$

$$-(2\sqrt{d-1} + \epsilon) \leq \lambda_n$$

( $\lambda_2 \leq d$ )

( $-d \leq \lambda_n$ )

( $d=2, 2\sqrt{d-1}=2$ )

( $d=3, 2\sqrt{d-1}=2\sqrt{2} < 3$ )

⋮

$$\rho(G) := \max_{i \geq 2} |\lambda_i|$$

$$= \max(\lambda_2, -\lambda_n)$$

↘  $\rho(G) \leq 2\sqrt{d-1} + \epsilon$

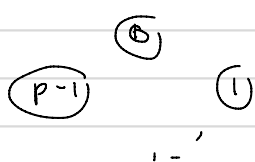
Upshot:  $G$ :  $\rho(G) := \max_{i \geq 2} |\lambda_i(\text{Adjacency}_G)|$

$$\textcircled{1} \left| e(U, W) - \frac{d}{n} |U| |W| \right| \leq \rho \sqrt{\frac{|U|(n-|U|)}{n}} \sqrt{\frac{|W|(n-|W|)}{n}}$$

$$\leq \rho \sqrt{|U|} \sqrt{|W|}$$

"expander mixing lemma"

$\textcircled{2}$  If  $p \equiv 1 \pmod{4}$



Cayley  $(\mathbb{Z}/p\mathbb{Z}, \{\text{quad residues}\})$

$n=p$ ,  $d = \frac{p-1}{2}$ ,  $\rho \approx$  better than  $2\sqrt{d-1}$

$\textcircled{3}$  A few more examples... General tools ...



To prove ①:

$$\text{Adj}_G = \underbrace{\frac{d}{n} \begin{bmatrix} 1 & \dots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \dots & 1 \end{bmatrix}}_{\substack{d \left( \text{proj}_{\vec{1}} \right) \\ \left( \frac{\vec{1}}{\sqrt{n}} \right) \left( \frac{\vec{1}}{\sqrt{n}} \right)^T}} + \underbrace{\varepsilon}_{\substack{\text{Adj}_G \Big|_{\vec{1}^\perp} \\ \text{norm in } L^2 \text{ of sense } = \rho}}$$

=

$$\begin{aligned} A_{\text{sym}} &= \sum_{i=1}^n \lambda_i \underbrace{\vec{u}_i \vec{u}_i^T}_{\text{for orthonormal eigenbasis } \vec{u}_1, \dots, \vec{u}_n} \\ &= \sum_{i=1}^n \lambda_i \text{proj}_{\vec{u}_i} \\ &= \lambda_1 \vec{u}_1 \vec{u}_1^T + \sum_{i=2}^n \lambda_i \text{proj}_{\vec{u}_i} \end{aligned}$$

=

eg.

$$A = \begin{bmatrix} 1 & \dots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \dots & 1 \end{bmatrix} - I = \text{Adj}_{\text{complete graph}} \quad d = \lambda = n-1, -1, -1, \dots, -1$$

$$= \frac{n}{n} \begin{bmatrix} 1 & \dots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \dots & 1 \end{bmatrix} + (-1) \text{proj}_{\vec{1}^\perp}$$

Jan 21:

$$D = \begin{bmatrix} d_1 & & & 0 \\ & d_2 & & \\ & & \ddots & \\ 0 & & & d_n \end{bmatrix} = \underbrace{\begin{bmatrix} d_1 & 0 & & \\ & \ddots & & \\ & & \ddots & \\ & & & \ddots \end{bmatrix}}_{D|_{\vec{e}_1}} + \underbrace{\begin{bmatrix} 0 & & & \\ & d_2 & & \\ & & d_3 & \\ & & & \ddots \end{bmatrix}}_{D|_{\text{span}(\vec{e}_2, \dots, \vec{e}_n)}} = \text{etc.}$$

$D|_{\vec{e}_1}$

$D|_{\text{span}(\vec{e}_2, \dots, \vec{e}_n)}$

$\{\vec{e}_1\}^\perp$

$$\vec{e}_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \vec{e}_2 = \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \dots, \vec{e}_n = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix}$$

$$D\vec{e}_i = d_i \vec{e}_i$$

$$D = \underbrace{\begin{bmatrix} d_1 & & & \\ & d_2 & & \\ & & 0 & \\ & & & d_4 & \\ & & & & \ddots \end{bmatrix}}_{\text{span}(\vec{e}_1, \vec{e}_2, \vec{e}_4)} + \underbrace{\begin{bmatrix} 0 & & & \\ & d_3 & & \\ & & c & \\ & & & d_5 & \\ & & & & d_6 & \\ & & & & & \ddots \end{bmatrix}}_{\text{span}(\vec{e}_3, \vec{e}_5, \dots, \vec{e}_n)}$$

Really  
 $d_1 = \lambda_1$   
 $d_2 = \lambda_2$   
 $\vdots$   
 in this case



$\text{span}(\vec{e}_1, \vec{e}_2, \vec{e}_4)$

$\text{span}(\vec{e}_3, \vec{e}_5, \dots, \vec{e}_n)$

Why works:

$D|_{\text{span}(\vec{e}_1, \vec{e}_2, \vec{e}_4)}$   
 $S_1$

image  $\subset \text{span}(\vec{e}_1, \vec{e}_2, \vec{e}_4)$   
 $S_2$

$$D: S_1 \rightarrow S_1$$

$$\mathbb{R}^n = \text{span}(S_1, S_2)$$

$$D: S_2 \rightarrow S_2$$

$$= S_1 \oplus S_2$$

Similarly for

$$D = \begin{bmatrix} d_1 & & & \\ & 0 & & \\ & & \ddots & \\ & & & \ddots \end{bmatrix} + \begin{bmatrix} 0 & & & \\ & d_2 & & \\ & & c & \\ & & & \ddots \end{bmatrix} + \begin{bmatrix} 0 & & & \\ & d_3 & & \\ & & c & \\ & & & \ddots \end{bmatrix} + \dots$$

RETURN TO THIS LATER

Difficult open problems:

—  
Question: For fixed  $d$  and  $n$  large  $\left( \begin{array}{l} n < d^2 \text{ or so} \\ \text{uninteresting} \end{array} \right)$

does there exist a  $d$ -reg graph on

$n$  vertices with  $\rho \leq 2\sqrt{d-1}$ ? Can you "construct" one?

—  
Question: For fixed  $d$  and  $n$  large, for how large a

$k=k(d)$  can you "construct" a graph such that

if  $|A|, |B| \geq \frac{n}{k}$ ,  $\#E(A, B) \geq 1$ ?

—  
Question: Fix any  $d \geq 3$ . Is it true that as  $n \rightarrow \infty$ ,

for all  $d$ -regular graphs,  $G$ , on  $n$  vertices,

$$\text{girth}(G) \leq (2 - \varepsilon) \log_{d-1} n$$

for some  $\varepsilon > 0$ .

Graph:  $U \subset \bar{V} \leftarrow$  vertices

$$\chi_U = \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \\ \vdots \\ 0 \\ 1 \\ 0 \end{bmatrix} = \begin{cases} \text{the } v^{\text{th}} \text{ component of} \\ U \text{ is } 1 \text{ if } v \in U \\ \text{otherwise } 0 \end{cases}$$

$\mathbb{R}^{\bar{V}}$

$\mathbb{R}^n$ ,

where  
 $n = |V|$

Rem:  $U, W \subset \bar{V} \leftarrow$  vertices of a graph

$$\begin{aligned} \chi_U^T (\text{Adj of } G) \chi_W &= [1 \ 0 \ 1 \ 0 \dots] \begin{bmatrix} \text{Adj} \end{bmatrix} \begin{bmatrix} 1 \\ \vdots \\ 0 \end{bmatrix} \\ &= e(U, W) \end{aligned}$$

=

$$\text{But: } \mathbb{R}^{\bar{V}} = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} \oplus \left\{ \begin{bmatrix} \vdots \\ \vdots \end{bmatrix} \right\}^{\perp}$$

$$\begin{aligned} \text{proj}_{\begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}}(\chi_U) &= \left( (\chi_U) \cdot \begin{bmatrix} 1/\sqrt{n} \\ \vdots \\ 1/\sqrt{n} \end{bmatrix} \right) \begin{bmatrix} 1/\sqrt{n} \\ \vdots \\ 1/\sqrt{n} \end{bmatrix} = \frac{|U|}{\sqrt{n}} \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} \\ &= \frac{|U|}{n} \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} = \frac{|U|}{n} \mathbf{1} \end{aligned}$$

$$\text{proj}_{\begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}^{\perp}}(\chi_U) = \chi_U - \text{proj}_{\begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}} \chi_U = \chi_U - \frac{|U|}{n} \mathbf{1}$$

$$\left| \underbrace{\text{proj}_{\begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix}^{\perp}}(\chi_U)} \right| = \text{magic} \sqrt{\frac{|U|(n-|U|)}{n}}$$

is this intuitive  $\begin{cases} U = \emptyset \checkmark \\ U = \bar{V} \end{cases}$

General Facts:

IF  $A$  is symmetric,

$S_1 =$  span of some eigenvect in ON eigenbasis

$$S_2 = S_1^\perp$$

then

$$A|_{S_1} : S_1 \rightarrow S_1$$

both symmetric

$$A|_{S_2} : S_2 \rightarrow S_2$$

$$A = \begin{bmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \\ & & & \lambda_n \end{bmatrix}$$

wrt a "nice" orthonormal basis

Fact: Say  $A$  sym, eigs  $\lambda_1 \geq \dots \geq \lambda_n$

Then

$$\|A\|_{L^2\text{-operator}} = \max_{\vec{v} \neq 0} \frac{\|A\vec{v}\|_2}{\|\vec{v}\|_2} = \max_{i=1, \dots, n} |\lambda_i|$$

Pf! Say that

Case 1:  $A = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$ ,  $\vec{v} = \sum c_i \vec{e}_i$

by scaling  $\vec{v}$ : can assume  $\sum c_i^2 = 1$

Then

$$A\vec{v} = A \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} = \begin{bmatrix} \lambda_1 c_1 \\ \vdots \\ \lambda_n c_n \end{bmatrix}, \text{ norm}^2 = \sum \lambda_i^2 c_i^2$$

if  $\vec{v} = \vec{e}_j$ ,  $\text{norm}^2 = \lambda_j^2$