

CPSC 536F

March 10, 2022

Last time: really used

$$e(u, w) = \frac{d}{n} |u| |w| \pm \rho \sqrt{|u| |w|}$$

really took $A_G \rightarrow (A_G)^k$

$$\left(\begin{array}{l} \# \text{ walks length} \\ k \text{ from } u \text{ to } w \end{array} - \frac{d^k}{n} |u| |w| \right)$$

$$\leq \rho^k \sqrt{|u| |w|}$$

u, w single vertices, $|u| = |w| = 1$

\Rightarrow # walks from any vertex to any other length k

$$\geq \frac{d^k}{n} - p^k$$

this is > 0 when

$$\frac{d^k}{n} > p^k$$

so

$$\frac{d^k}{p^k} > n \quad \text{so} \quad k > \frac{\log n}{\log(d/p)}$$

Gives:

diameter of G

$$\stackrel{\text{def}}{=} \max_{v_1, v_2 \in V_G} (\text{distance from } v_1 \text{ to } v_2)$$

$$v_1, v_2 \in V_G$$

$$\leq \frac{\log n}{\log(d/p)} + 1$$

Today start!

(1) If G is d -regular on n vertices, d fixed, $n \rightarrow \infty$

Alon-
Beppana $\lambda_2(G) \geq 2\sqrt{d-1} - o_n(1)$

specifically

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

(2) If $d \geq 4$, even, fixed, then most d -regular graphs on n vertices have $\lambda_2(G) \leq 2\sqrt{d-1} + \epsilon$ for any fixed ϵ

$$\lambda_2(G) \leq 2\sqrt{d-1} + \epsilon \quad [Me]$$

first result

$$\lambda_2(G) \geq d^{1/2} (2\sqrt{d-1})^{1/2} + \epsilon$$

(Broder-Shamir)

↓
Used Adjacency G

Using

{ non-backtracking } matrix
{ Washimoto }

get a better result.

=

Why $2\sqrt{d-1}$...

If G is any graph, A_G
symmetric, not necessarily regular,
then

$$\lambda_1(A_G) \geq \lambda_2 \geq \dots \geq \lambda_n(G)$$

$\log_2(\lambda_1(A_G)) :=$ Shannon
capacity
of G

e.g.

Look at strings 1's, 0's s.t.

no 00 occurs as a substring

"Fibonacci data"

strings length 3!

111		
011		001
101	not	100
110		000
010		

=

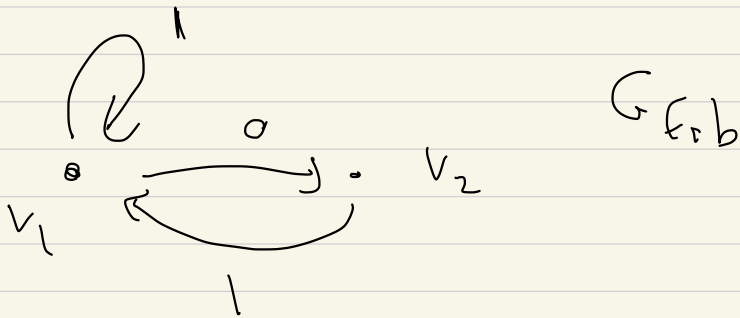
Similarly look at strings 0,1's
sit. between two 1's,

at least d 0's occur

at most k 0's occur

" (d, k) - constrained data"

Fibonacci data:

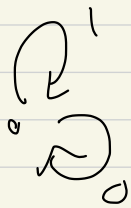


all Fibonacci data can be obtained by walks on this

graph

start at v_1 end v_1
 v_2

=



$$\text{Adj } G = [2]$$

← models strings 0, 1 unconstrained

Shannon's theory:

Say you have a directed graph

words of length k \sim $\binom{k}{\lambda_1}$

↑
hiding constants depend on graph

Let's make this precise ...

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Thm: Say that digraph G that is strongly connected

i.e. for any $v_1, v_2 \in G$

there is a path from v_1 to v_2

[Mirrors the notion of an irreducible Markov chain.]

Then there is a single largest positive eigenvalue λ_1 in absolute value, and all other eigenvalues have absolute value $\leq \lambda_1$.

=

If G is a graph:

$$\lambda_1(A_G) \geq \dots \geq \lambda_n(A_G)$$

↑
positive

↑
no smaller than
 $-\lambda_1(A_G)$

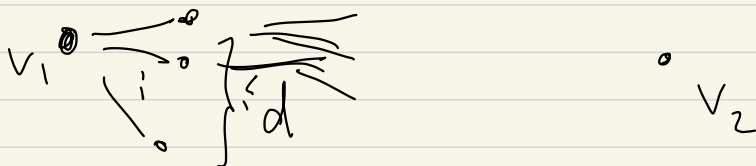
Take any d -regular graph G



Want to show $\lambda_2(G) \geq 2\sqrt{d-1}$
+ small

Strategy:

take any vertex $v_1 \in V_G$. Take
a v_2 s.t. distance v_1 to v_2
is as large as possible



$$\leq d(d-1)$$

vertices dist 1 to $v_1 \leq d$

" " 2 to $v_1 \leq d(d-1)$

3 to $v_1 \leq d(d-1)^2$

if

$$f(k) = 1 + d + d(d-1) + d(d-1)^2 + \dots + d(d-1)^{k-1}$$

$$< n$$

then some vertex has distance

$$\geq k.$$

$$\text{So } f(k) = d(d-1)^k \left(1 + \frac{1}{d-1} + \dots + \frac{1}{(d-1)^{k-1}} \right) + 1 + d$$

So if $d \geq 3$, $1 + \frac{1}{d-1} + \frac{1}{(d-1)^2} + \dots \leq 2$

$$d(d-1)^k \geq n - d - 1$$

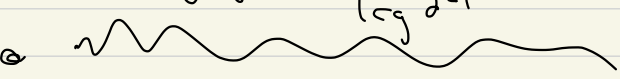
there is a v_2 at distance $k+1$ to v_1 .

$$\text{So take } k \geq \frac{\log\left(\frac{n-d-1}{2d}\right)}{\log(d-1)} - 1$$

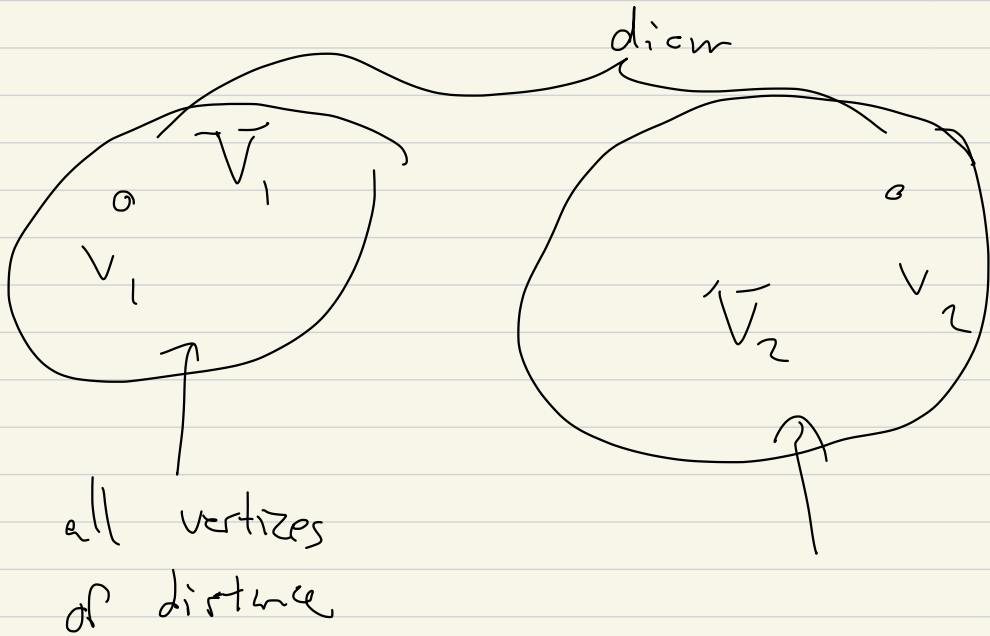
$$\geq \frac{\log n}{\log(d-1)} - c'$$

So

$$\text{diam} = \text{dist} \left(\frac{\log n}{\log d-1} - c' \right)$$

v_1  v_2

If $\text{diam} =$ diameter of G



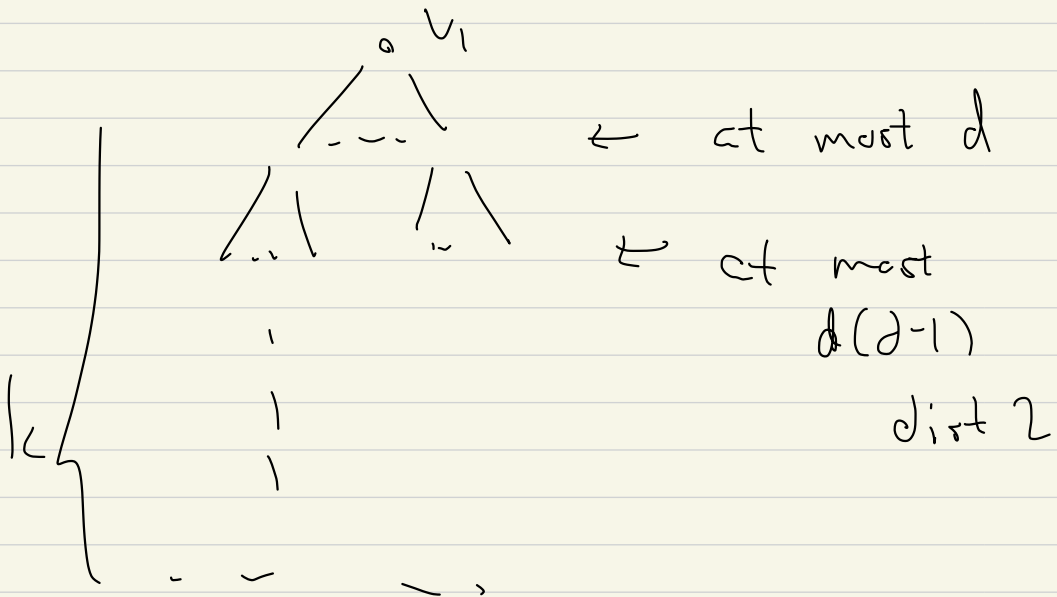
$$\leq \frac{\text{diam}}{2} - 3$$

$\bar{V}_i =$ set of vertices of distance

$$\leq \frac{\text{diam}}{2} - 3 \text{ from } v_i$$

Claim: d -regular graph G ,
 and any subgraph of vertices
 of distance k from a

$$v_1 \in V$$



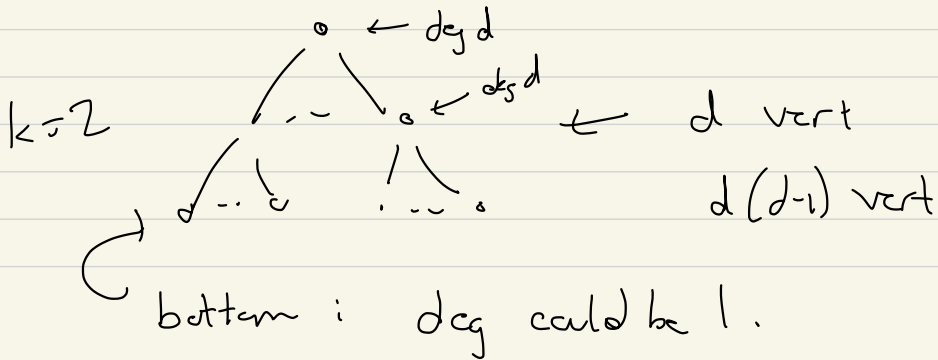
Then

λ_1 (Induced subgraph
 or \bar{V}_1 by
 G)

$$\geq 2\sqrt{d-1} - \underbrace{\text{function}(k)}_{\rightarrow 0 \text{ as } k \rightarrow \infty}$$

=

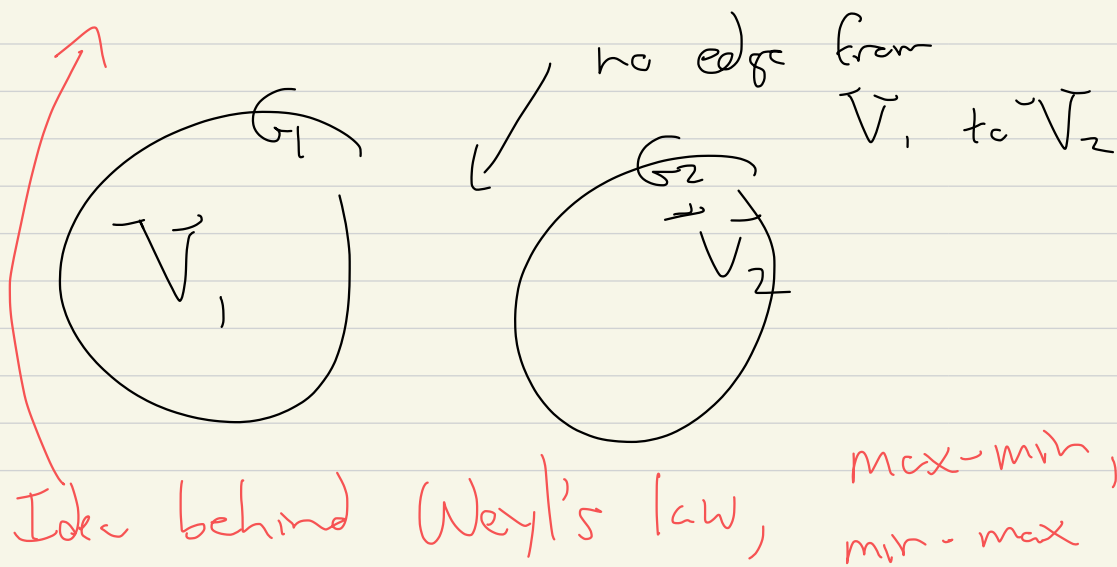
e.g. if G is locally a tree



Thm! If G_1 is the induced graph
 by G taking a subset V_1
 as above,

$$\lambda_1(G_1) \geq 2\sqrt{d-1} + o_{1 \rightarrow 2}(1)$$

Thm! If you have



Then

$$\lambda_2(G) \geq$$

$$\min(\lambda_1(G_1), \lambda_1(G_2))$$

= Similarly

$$\lambda_1(G_1)$$

$$\lambda_1(G_2)$$

$$\lambda_1(G_3)$$

$$\lambda_3(G) \geq \min(\lambda_1(G_1), \lambda_1(G_2), \lambda_1(G_3))$$

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Min-max & max-min principles!

Say you have symmetric $n \times n$ matrix, A , eigenvalues

$\lambda_1, \dots, \lambda_n$

Rayleigh quotient: $\vec{v} \neq 0, \vec{v} \in \mathbb{R}^n$

$$\mathcal{R}(\vec{v}) = \frac{(A\vec{v}) \cdot \vec{v}}{\vec{v} \cdot \vec{v}}$$

Rem: If $\alpha \neq 0, \alpha \in \mathbb{R}$

$$\mathcal{R}(\alpha\vec{v}) = \frac{(A\vec{v}\alpha) \cdot (\vec{v}\alpha)}{(\vec{v}\alpha) \cdot \vec{v}\alpha} = \mathcal{R}(\vec{v})$$

Let $\vec{v}_1, \dots, \vec{v}_n$ be orthonormal
eigenbasis, $A\vec{v}_i = \lambda_i \vec{v}_i$

Then if $\vec{v} \in \mathbb{R}^n$

$$\vec{v} = c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_n \vec{v}_n$$

where

$$c_i = \vec{v} \cdot \vec{v}_i$$

$$\vec{v} \cdot \vec{v} = (c_1 \vec{v}_1 + \dots + c_n \vec{v}_n) \cdot (c_1 \vec{v}_1 + \dots + c_n \vec{v}_n)$$

$$= c_1^2 + c_2^2 + \dots + c_n^2$$

(Pythagoras' law)

$$R(\vec{v}) = \frac{\left(A(c_1 \vec{v}_1 + \dots + c_n \vec{v}_n) \right) \cdot (c_1 \vec{v}_1 + \dots + c_n \vec{v}_n)}{\vec{v} \cdot \vec{v}}$$

$$= \frac{(c_1 \vec{v}_1 \cdot \lambda_1 + \dots + c_n \vec{v}_n \cdot \lambda_n)}{c_1^2 + c_2^2 + \dots + c_n^2}$$

$$= \frac{\lambda_1 c_1^2 + \lambda_2 c_2^2 + \dots + \lambda_n c_n^2}{c_1^2 + c_2^2 + \dots + c_n^2}$$

$$= \lambda_1 \left(\frac{c_1^2}{c_1^2 + \dots + c_n^2} \right) + \lambda_2 \left(\frac{c_2^2}{c_1^2 + \dots + c_n^2} \right) + \dots$$

For any $\vec{v} \neq 0$

Rem: $R(\vec{v}) \leq \lambda_1$,

Since

$$R(\vec{v}) = \lambda_1 \left(\begin{array}{c} \leftarrow \text{non-neg reals, sum 1} \\ \downarrow \\ \dots \end{array} \right) + \dots + \lambda_n \left(\begin{array}{c} \leftarrow \text{non-neg reals, sum 1} \\ \downarrow \\ \dots \end{array} \right)$$

$$\leq \lambda_1 \left(\begin{array}{c} \leftarrow \text{non-neg reals, sum 1} \\ \downarrow \\ \dots \end{array} \right) + \dots + \lambda_n \left(\begin{array}{c} \leftarrow \text{non-neg reals, sum 1} \\ \downarrow \\ \dots \end{array} \right)$$

$$= \lambda_1$$

and $R(\vec{v}_1) = \lambda_1$

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If $W \subset \mathbb{R}^n$, $\dim(W) = 2$,

then some $\vec{w} \in W$, $\vec{w} \neq 0$ s.t.

$$\mathcal{R}(\vec{w}) \leq \lambda_2.$$

Why: $\vec{W} = \text{span}(\vec{w}_1, \vec{w}_2)$

Claim: There is a non-zero

α_1, α_2 s.t.

$$(\alpha_1 \vec{w}_1 + \alpha_2 \vec{w}_2) \cdot \vec{v}_1 = 0$$

Why?

$$\alpha_1 (\vec{w}_1 \cdot \vec{v}_1) + \alpha_2 (\vec{w}_2 \cdot \vec{v}_1) = 0$$

fact! Any linear system on m variables and $m-1$ equations has

a non-trivial solution.

So there are α_1, α_2 , not all 0
s.t.,

$$\vec{w} = \alpha_1 \vec{w}_1 + \alpha_2 \vec{w}_2 \neq \vec{0}$$

is orthogonal to ~~\vec{w}_1~~ , . So

$$\vec{w} = c_1 \vec{v}_1 + c_2 \vec{v}_2 + \dots + c_n \vec{v}_n$$

$$c_1 = \vec{w} \cdot \vec{v}_1 = 0$$

Hence

$$R(\vec{w}) = \lambda_1 \left(\frac{c_1^2}{\sum c_i^2} \right) + \lambda_2 \left(\frac{c_2^2}{\sum c_i^2} \right) + \dots$$

$$= \lambda_2 \left(\frac{c_2^2}{\sum_i c_i^2} \right) + \dots + \lambda_n \left(\frac{c_n^2}{\sum_i c_i^2} \right)$$

$$\leq \lambda_2 () + \dots + \lambda_2 ()$$

$$\leq \lambda_2.$$

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So $\dim(W) = 2$ (or ≥ 2)

then some $\vec{w} \in W$, $\vec{w} \neq 0$ has

$\vec{w} \perp \vec{v}_1$, hence

$$R(\vec{w}) \leq \lambda_2$$

Gives lower bound on λ_2 .

Also, if $W = \text{Span}(\vec{v}_1, \vec{v}_2)$

then $\forall \vec{w} \in W$,

$$\vec{w} = c_1 \vec{v}_1 + c_2 \vec{v}_2$$

$$\mathcal{R}_A(\vec{w}) = \frac{c_1^2 \lambda_1 + c_2^2 \lambda_2}{c_1^2 + c_2^2}$$

$$\geq \frac{c_1^2 \lambda_2 + c_2^2 \lambda_2}{c_1^2 + c_2^2} = \lambda_2$$

=

$$\max_{\substack{\dim(W) = 2, W \subset \mathbb{R}^n \\ \vec{w} \in W \\ \vec{w} \neq \vec{0}}} \left(\min_{\vec{w} \in W} \mathcal{R}_A(\vec{w}) \right)$$

$$= \lambda_2$$

max - min principle.

=

In particular $\dim(W) = 2$

\Rightarrow

$$\lambda_2 \geq \min_{\substack{\vec{w} \in W \\ \vec{w} \neq 0}} R(\vec{w})$$

idea

