

April 1, 2021, CPSC 531F

- Today:

- 3<sup>rd</sup> Proof that symmetric matrices have ON eigenbasis, but more generally!  $M \in \mathbb{M}_n(\mathbb{C})$ , say that  $M$  is "normal" if

$$M^H M = M M^H$$

( symmetric, Hermitian, orthogonal, unitary, ... )

all have ON eigenbasis

"Schur decomposition"

- Perron-Frobenius thm :

If  $A \in M_n(\mathbb{R})$  and has  
non-negative real entries,  
and all rows have at least one  
positive entry !

$\lambda$  Perron-Frobenius eigenvalue ,

if  $A$  is irreducible, then

a bunch of nice things happen ...

Lemma: (Schur decomposition)

Any  $A \in M_n(\mathbb{C})$  can be

written as

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

some stuff above the diagonal

(where  $U$  is invertible)

$\lambda_1, \dots, \lambda_n$  are the eigenvalues of  $A$

$$U = \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \\ | & & | \\ 1 & & 1 \end{bmatrix}, \quad \vec{v}_1, \dots, \vec{v}_n \text{ are ON as vectors in } \mathbb{C}^n$$

So

$$AU = UT$$

(also  
 $A = UTU^{-1}$ )

T is  $\begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$  is "upper

triangular."

$$U^H U = \begin{bmatrix} -\vec{v}_1^H & & \\ & \ddots & \\ -\vec{v}_n^H & & \end{bmatrix} \begin{bmatrix} \vec{v}_1 & & \\ & \ddots & \\ & & \vec{v}_n \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & \\ & \ddots & 0 \\ 0 & & 1 \end{bmatrix} = I$$

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

for  $M_n(\mathbb{R})$  "orthogonal"

Say that  $U \in M_n(\mathbb{C})$  is unitary

if (1)  $U^H U = I$

(2)  $U = \begin{bmatrix} | & & | \\ \vec{v}_1 & \dots & \vec{v}_n \\ | & & | \end{bmatrix}$  with

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

(3)  $U^H U = I = U U^H$

(4)  $U = \begin{bmatrix} \text{---} \vec{u}_1^H \text{---} \\ | \\ | \\ \text{---} \vec{u}_n^H \text{---} \end{bmatrix}$

$\vec{u}_1, \dots, \vec{u}_n$  are ON basis

Proof (of Schur decomp):

$A \in M_n(\mathbb{C})$  has an eigenvalue  $\lambda$ ,  
corresponding eigenvector  $\vec{v}$ ,

$$A\vec{v} = \lambda\vec{v}. \quad \text{Write } \lambda_1 = \lambda, \quad \vec{v}_1 = \vec{v}/\|\vec{v}\|$$

[We know this  $\text{char poly}_A(\lambda) = \det(\lambda I - A)$ ,  
---  $\Rightarrow \lambda, \Rightarrow \vec{v} \in \text{Ker}(\lambda I - A)$ ]

[Theory, not algorithm. That's practical...]

Extend  $\vec{v}_1$  to an ON basis of  $\mathbb{C}^n$

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

Given  $\vec{v}_1 = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \\ i/\sqrt{2} \end{bmatrix}$ ,

(via Gram-Schmidt) ( $n = 10^3, 10^6$  this may not be so practical)

$$A \underbrace{\begin{bmatrix} | & | & \dots \\ \vec{v}_1 & \vec{v}_2 & \dots \\ | & | & \dots \end{bmatrix}}_{\text{Unitary}} = \begin{bmatrix} | & | & \dots \\ \vec{v}_1 & \vec{v}_2 & \dots \\ | & | & \dots \end{bmatrix} \begin{bmatrix} \lambda & ? & ? & ? \\ 0 & ? & ? & ? \\ 0 & ? & ? & ? \\ 0 & ? & ? & ? \\ 0 & ? & ? & ? \end{bmatrix}$$

?

$$A \vec{v}_1 = \lambda_1 \vec{v}_1$$

$A \vec{v}_2 = \text{some combo of } \vec{v}_1, \dots, \vec{v}_n$

$A \vec{v}_3 = \dots$

Now:

$$A = A_1$$

$$A U_1 = U_1 \begin{pmatrix} \lambda_1 & \begin{matrix} ? & ? & ? & ? & ? \end{matrix} \\ \hline 0 & \begin{matrix} \text{matrix} \\ \text{matrix} \\ \text{matrix} \\ \text{matrix} \\ \text{matrix} \end{matrix} \end{pmatrix}$$

(n-1) x (n-1)

now use induction / recursion

$$A_2 \text{ is } A U_2 = U_2 \begin{pmatrix} \lambda_2 & \begin{matrix} ? & ? & ? & ? \end{matrix} \\ \hline 0 & \begin{matrix} ? & ? & ? \\ ? & ? & ? \\ ? & ? & ? \end{matrix} \end{pmatrix}$$

$$U_1^{-1} A U_1 = \begin{pmatrix} \lambda_1 & \begin{matrix} ? & ? & ? & ? & ? \end{matrix} \\ \hline 0 & \begin{matrix} \text{matrix} \\ \text{matrix} \\ \text{matrix} \\ \text{matrix} \\ \text{matrix} \end{matrix} \end{pmatrix}$$



$$U_1^{-1} A U_1 = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & \vdots & & & \\ \vdots & & U_2 & & \\ 0 & & & & \end{bmatrix} = \begin{bmatrix} \lambda_1 & ? & ? & \dots & ? \\ 0 & \lambda_2 & ? & \dots & ? \\ \vdots & & \vdots & & \vdots \\ \vdots & & & & \vdots \\ \vdots & & & & \vdots \end{bmatrix}$$

$\begin{bmatrix} 1 & 0 & \dots \\ 0 & \vdots & \\ \vdots & & U_2 \\ 0 & & \end{bmatrix}$ 
 $\uparrow$   
 $A_3$

keep going.

$$n = 10^8$$

$$U_1^{-1} A U_1 = \begin{bmatrix} \lambda_1 & ? & ? & \dots & ? \\ 0 & \vdots & & & \\ \vdots & & U_2 & & \\ 0 & & & & \end{bmatrix} \leftarrow A_2$$

$$A_2 = \begin{bmatrix} \rightarrow & \rightarrow \\ \rightarrow & \rightarrow \\ \rightarrow & \rightarrow \\ \rightarrow & \rightarrow \end{bmatrix} \rightarrow A_2 U_2 = U_2 \begin{bmatrix} \lambda_2 & ? & ? & \dots & ? \\ \vdots & & \vdots & & \vdots \\ \vdots & & & & \vdots \\ \vdots & & & & \vdots \\ \vdots & & & & \vdots \end{bmatrix}$$

$\uparrow$   
 $A_3$

$$\dots \begin{pmatrix} | & | \\ \hline & \\ \hline | & | \\ \hline u_1 & u_2 \end{pmatrix} A \begin{pmatrix} | & | \\ \hline & \\ \hline | & | \\ \hline u_1 & u_2 \end{pmatrix} \begin{pmatrix} | & | \\ \hline & \\ \hline | & | \\ \hline u_2 & u_3 \end{pmatrix} \dots$$

gives you

$$\begin{pmatrix} \lambda_1 & & & & \\ & \ddots & & & \\ & & \ddots & & \\ & & & \ddots & \\ 0 & & & & \lambda_n \end{pmatrix} \Rightarrow \Gamma$$

$$\rightarrow U^{-1} A U = \Gamma$$

Algorithmically this is not so good ---

$$\text{If } A \begin{bmatrix} | & & | \\ v_1 & \dots & v_n \\ | & & | \end{bmatrix} = \begin{bmatrix} | & & | \\ v_1 & v_2 & \dots & v_n \\ | & & & | \end{bmatrix}^T$$

$$U^{-1}AU = \begin{bmatrix} t_{11} & t_{12} & t_{13} \\ 0 & t_{22} & t_{23} \\ 0 & 0 & t_{33} \\ \vdots & \vdots & \vdots \\ 0 & 0 & 0 & t_{nn} \end{bmatrix} = T$$

then if

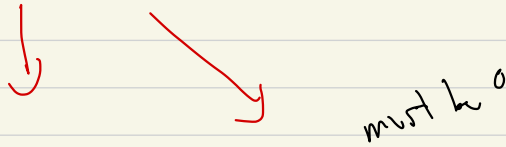
$$A^H A = A A^H, \text{ i.e. } A \text{ is normal}$$

then

$$T^H T = (U^{-1}AU)^H (U^{-1}AU)$$

$$\begin{aligned} (U^{-1} = U^H) & \quad (U^H A^H U) (U^{-1}AU) \\ & = U^{-1} A^H A U \end{aligned}$$

$$T^H T = T T^H \quad \text{then } T \text{ is also Normal.}$$



$$\begin{array}{l}
 \text{all 0} \\
 \left[ \begin{array}{cccc}
 \overline{t_{11}} & 0 & \dots & 0 \\
 \cancel{t_{12}} & \overline{t_{22}} & 0 & \dots & 0 \\
 \cancel{t_{21}} & \vdots & \vdots & \vdots & \vdots \\
 \vdots & \vdots & \vdots & \vdots & \vdots
 \end{array} \right]
 \end{array}
 \quad
 \begin{array}{l}
 \text{must be } 0 \\
 \left[ \begin{array}{cccc}
 t_{11} & \cancel{t_{12}} & \cancel{t_{13}} & \dots \\
 0 & t_{21} & \dots & \dots \\
 0 & 0 & \dots & \dots \\
 \vdots & \vdots & \vdots & \vdots \\
 \vdots & \vdots & \vdots & \vdots
 \end{array} \right]
 \end{array}$$

$$= \left[ \begin{array}{cccc}
 |t_{11}|^2 & & & \\
 & |t_{12}|^2 + |t_{22}|^2 & & \\
 & & \vdots & \\
 & & & \vdots
 \end{array} \right]$$

similarly  
 (2,2)  
 entry of  
 $T^H T = T T^H$

$T^H T$

$$\begin{pmatrix} t_{11} & t_{12} & \dots & t_{1n} \\ 0 & t_{22} & t_{23} & \dots \end{pmatrix} \begin{pmatrix} \bar{t}_{11} & 0 \\ \bar{t}_{12} & \bar{t}_{22} \\ \vdots & \bar{t}_{23} \\ \vdots & \vdots \end{pmatrix}$$

1st diagonal

$$= \left( \begin{array}{l} |t_{11}|^2 + |t_{12}|^2 + \dots \\ \dots \\ \dots \end{array} \right)$$

2,2 entry  $\Rightarrow t_{23} = 0 = t_{24} = \dots$

Look at 1,1 entry of  $T^H T = T T^H$

$$|t_{11}|^2 = |t_{11}|^2 + \underbrace{|t_{12}|^2 + \dots}_{\text{all 0}} \Rightarrow$$

So!

$T$  is upper tri, and normal

$$(1) T^H T = T T^H \Rightarrow T \text{ is diagonal}$$

$$(2) \text{ If } AU = UT, \quad U \text{ unitary,}$$

$$A^H A = A A^H \Rightarrow T \text{ is normal}$$

"Schur decomposition"

$\Rightarrow$  any normal  $A$  has ON eigenbasis

e.g.  $U^{-1} = U^H$ ,  $U^H U = I$

then

$$U^H U = I = U U^H \left. \vphantom{U^H U} \right\} \text{normal}$$

So  $U$  is normal.

$\Rightarrow$

$A$  is skew-Hermitian if

$$A^H = -A$$

$$A^H A = A A^H = -A^2$$

normal

# Perron-Frobenius!

Let  $A \in M_n(\mathbb{R})$  with non-neg  
entries

e.g.  $A$  Markov matrix

$A$  adj mat (di)graph

$A$  from (2,7)-codes  
information

---

We define for such  $A$ !

$\lambda_{\text{PF}}(A)$ :



(1)  $\lambda_{PF}$  is the largest real eigenvalue of  $A$

Then we assume  $A$  is irreducible, and prove

- any eigenvalue of  $A$  has absolute value at most  $\lambda_{PF}$

-  $\lambda_{PF}$  has multiplicity 1

-  $\lambda_{PF}$  has unique eigenvector,

$\vec{v}$ , up to scaling, and

we can take  $\vec{v} > \vec{0}$ ,  $\vec{v}$  to  
have all real, positive components...

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Let's define, for any  $A \in M_n(\mathbb{R})$   
with non-negative entries!

$$PF(A) = \left\{ \begin{array}{l} \lambda \in \mathbb{R} \\ \lambda \geq 0 \end{array} \right\} \left\{ \begin{array}{l} A\vec{v} \geq \lambda\vec{v} \text{ for} \\ \text{some } \vec{v} \geq \vec{0} \end{array} \right\}$$

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

so  $3 \in PF(A)$

We look at

$$\max \lambda \in \text{PF}(A)$$

and use  $\lambda_{\text{PF}}$  for this maximum value,

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 4 \\ 5 \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \end{bmatrix} \Rightarrow \begin{pmatrix} 4 \\ 5 \end{pmatrix} \begin{bmatrix} 4 \\ 5 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 3 \\ 7 \end{bmatrix} = \begin{bmatrix} 7 \\ 3 \end{bmatrix} \Rightarrow \begin{pmatrix} 3 \\ 7 \end{pmatrix} \begin{bmatrix} 3 \\ 7 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \end{bmatrix} \Rightarrow 1 \cdot \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

$$\begin{bmatrix} a & b \\ b & a \end{bmatrix} \quad \lambda = a+b, a-b$$

$$\begin{bmatrix} 3 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 2 \\ 7 \end{bmatrix} = \begin{bmatrix} 13 \\ 23 \end{bmatrix}$$

$$\approx \text{something} \begin{bmatrix} 2 \\ 7 \end{bmatrix}$$

↑

$$\begin{bmatrix} 3 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 6 \\ 7 \end{bmatrix} \approx \begin{bmatrix} \nearrow \\ \nearrow \end{bmatrix} \begin{bmatrix} 6 \\ 7 \end{bmatrix}$$

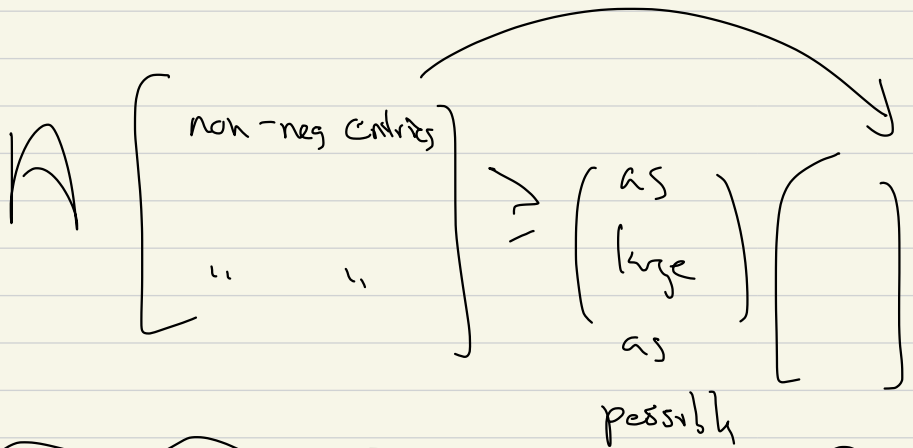
close to 4

$\frac{23}{7}$  not quite 4

$$\begin{pmatrix} 3 & 1 \\ \cancel{1} & 3 \end{pmatrix} \begin{pmatrix} 5 \\ 5 \end{pmatrix} = \begin{pmatrix} 20 \\ 20 \end{pmatrix} \Rightarrow 4 \cdot \begin{pmatrix} 5 \\ 5 \end{pmatrix}$$

$$\begin{pmatrix} 20 \\ 20 \end{pmatrix} = 4 \cdot \begin{pmatrix} 5 \\ 5 \end{pmatrix}$$

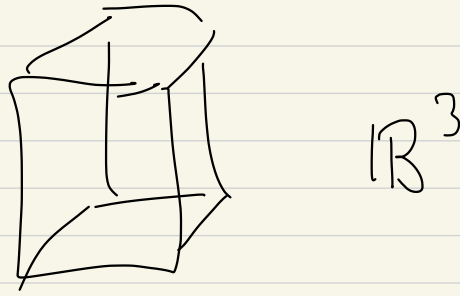
λ pf eigenvalue!



Class ends

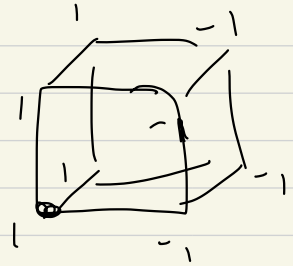
$G$  Cayley graph,

$$G = \mathbb{R}^n$$

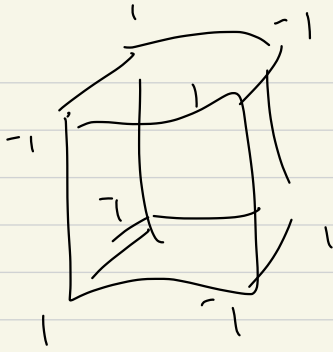


$G_{\text{graph}} = (\mathbb{Z}^3)^3 \leftarrow \text{group}$

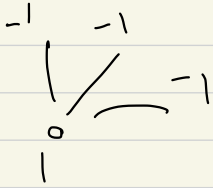
$A_G$  eigenvectors



$$\lambda = 1 \quad \begin{matrix} \text{---} \\ | \\ \bullet \\ | \\ \text{---} \end{matrix} \quad \begin{matrix} \text{---} \\ | \\ \bullet \\ | \\ \text{---} \end{matrix}$$



$$\lambda = -3$$



Eigenvectors with  $\pm 1$  values

$$\left( \begin{array}{c} \sqrt{2} \\ 1 \\ \sqrt{2} \end{array} \right)^3 \xrightarrow{\chi} \textcircled{1}$$

$$\chi(g_1 g_2) = \chi(g_1) \chi(g_2)$$

$G$  based on Abelian group,

$$f: V \rightarrow \mathbb{R}$$

$$f = \text{sum of } \sum (c_i) \vec{v}_i$$

↓  
eigenvectors

$f \rightarrow$  if  $\vec{v}_1, \dots, \vec{v}_n$  ON

$$f = \sum_{i=1}^n \underbrace{(f \cdot \vec{v}_i)}_{\rightarrow \text{"Fourier coeffs of } f"} \vec{v}_i$$

$$\hat{f}(x) := (f \cdot x) \frac{1}{\sqrt{n}}$$



$G$  is Abelian, then a character  
of  $G$  is map

$$\chi : G \rightarrow \mathbb{C}^{\text{non-zero}}$$

$$\chi(g_1 g_2) = \chi(g_1) \chi(g_2)$$

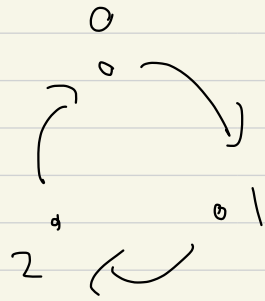
$$G = \mathbb{R} :$$

$$(2\pi i g) \cdot g$$

$$\chi(g) = e$$

$\chi$  has values on unit circle

$$C_3 = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix}$$



modulo 3,

$$g = \text{group} = \mathbb{Z} / 3\mathbb{Z}$$

eigen vectors

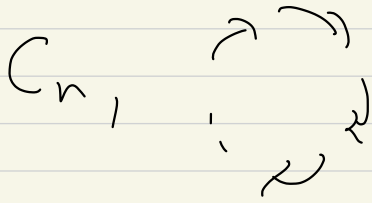
$$\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} : g \mapsto 1^g$$

$$\begin{pmatrix} 1 \\ \omega \\ \omega^2 \end{pmatrix} : g \mapsto \omega^g$$



$$\begin{pmatrix} 1 \\ \omega^2 \\ \omega^4 \end{pmatrix} : g \mapsto (\omega^2)^g$$

$$\omega = e^{2\pi i / 3}$$



group  $\mathbb{Z}/n\mathbb{Z}$

$$\sum^n = 1$$

$$\chi_3(g) \mapsto \sum^g$$

Fourier  
coefs

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = c_1 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ \omega \\ \omega^2 \end{bmatrix} + c_3 \begin{bmatrix} 1 \\ \omega^2 \\ \omega \end{bmatrix}$$

Fourier trans  $\begin{bmatrix} x \\ y \\ z \end{bmatrix} \rightarrow$

- $c_1 \leftarrow \text{char } g \mapsto 1^g$
- $c_2 \leftarrow g \mapsto \omega^g$
- $c_3 \leftarrow g \mapsto \omega^{2g}$

Poss?

3 present, ~25 min last day

April 13

One (or more)

April 8

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- Total Unimodularity

- Electrical Network

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Coupling:

$$\bar{d}(s+t) \leq \bar{d}(s) \bar{d}(t)$$

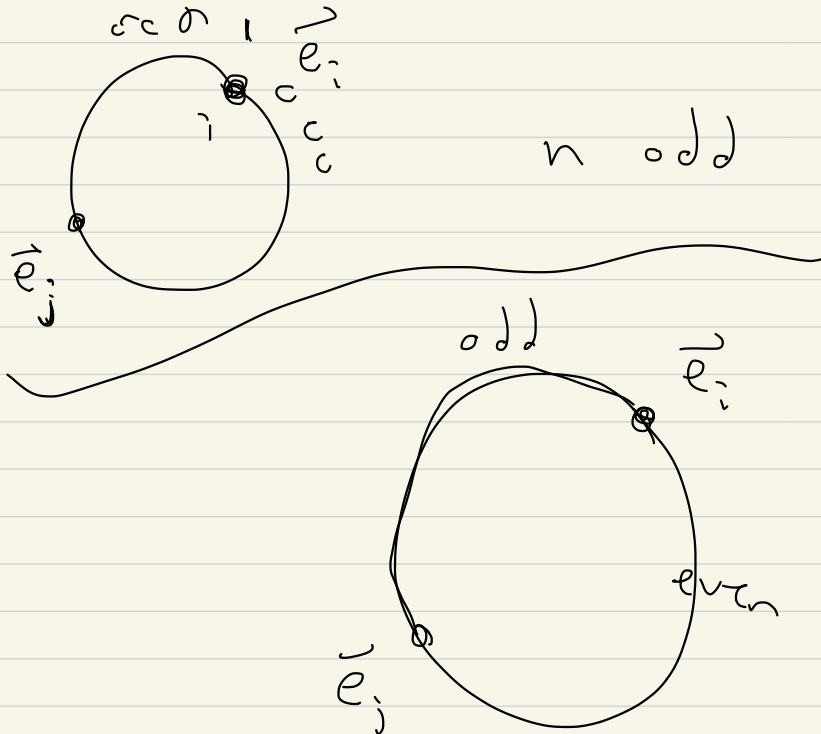
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4.11

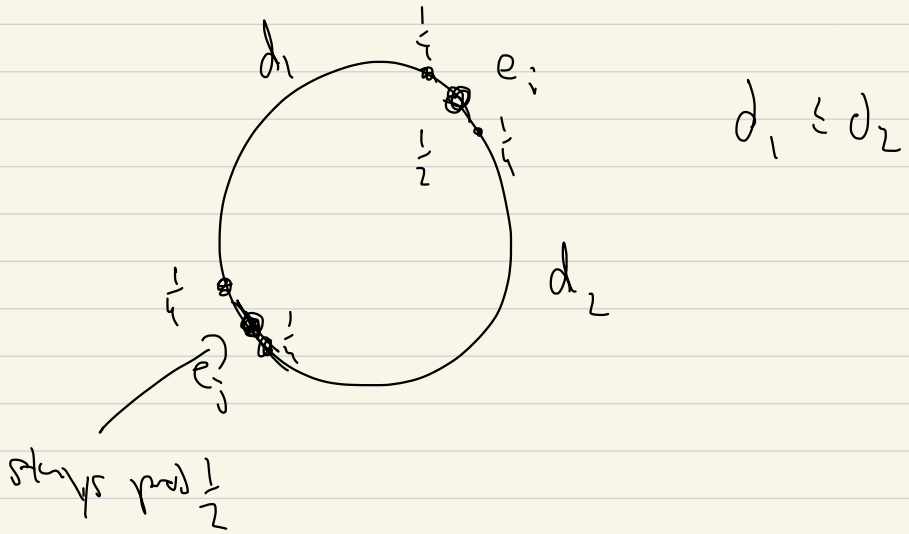
Another coupling

Coupling examples.

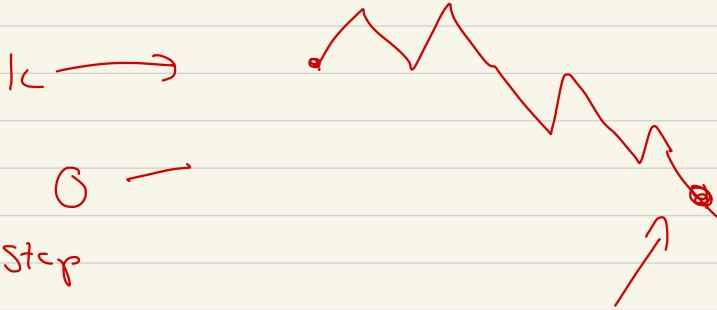
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lazy  $\frac{1}{2}$ , left or right  $\frac{1}{4}$



step  
n



$$E[\text{time to step}] = k(n-k)$$

$$\text{Prob}(\text{time to step} \geq t) = f$$

$$\leq E[\text{time to step}]$$

X takes on non-neg values

$$E[X] = 0 \cdot \text{Prob}(X=0) \\ + 1 \cdot \text{Prob}(X=1) \\ + \vdots$$

$$\text{any } t \geq t \cdot \text{Prob}(X=t) \\ + (t+1) \cdot \text{Prob}(X=t+1) \\ + \vdots$$

$$\geq t \cdot \text{Prob}(X=t) \\ + t \cdot \text{Prob}(X=t+1) \\ + \vdots$$

$$= t \cdot \text{Prob}(X \geq t)$$

---


$$\text{Prob}\left(\frac{\text{time to step}}{\text{step}} \geq t\right) \leq \frac{k(n-k)}{t}$$

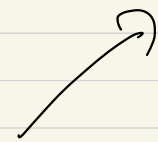


Gambler's ruin!

Prbd (time to step  $\geq t$ )

$\leq$

$$\frac{k \cdot (n-k)}{t} \quad \text{max all } k$$



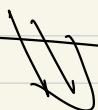
Gambler's  
ruin  
formula

$\leq$

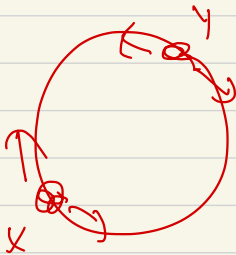
$$\frac{(n/2)(n - n/2)}{t}$$

$\approx$

$$\frac{n^2}{4t}$$



Mixing time



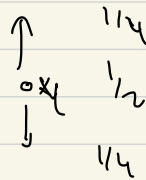
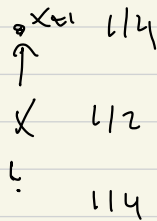
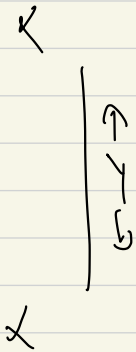
lazy!

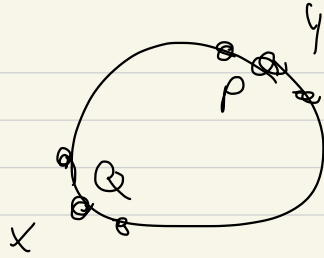
if  $x=y$ , either  $\left( \begin{array}{l} x \text{ moves} \\ y \text{ stays} \end{array} \right.$

or  $\left( \begin{array}{l} x \text{ lazy} \\ y \text{ moves} \end{array} \right.$

$$\text{Time for } \bigcup = X \leq \frac{n^2}{4t}$$

$$\max_{i,j} \left| \sum e_i p^t - \sum e_j p^t \right| \leq \text{time when gamblers run has prob } 1/4$$





$$U_{t+1} = P_{-1}, P, P_{+1}$$

$$X_{t+1} = Q_{-1}, Q, Q_{+1}$$

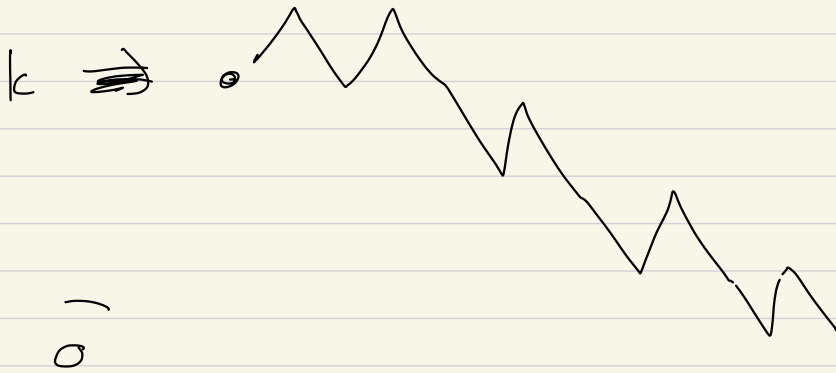
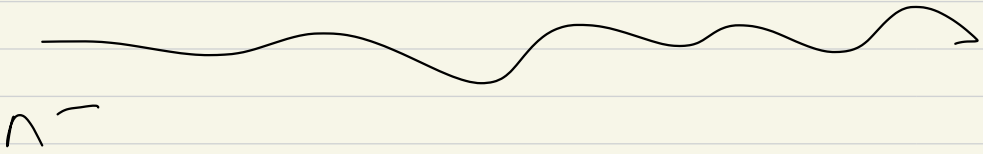
$$\left. \begin{aligned} \text{Prob}(P, Q) &= 1/4 \\ \text{Prob}(P, Q_{\pm 1}) &= 1/8 \\ \text{Prob}(P_{\pm 1}, Q_{\pm 1}) &= 1/6 \end{aligned} \right\} \begin{array}{l} \text{If } X, U \\ \text{independent} \end{array}$$

Max time:

$$\max_i \left| \vec{e}_i \cdot \vec{p}^t - \vec{e}_j \cdot \vec{p}^t \right|$$

$$\leq 1/4$$

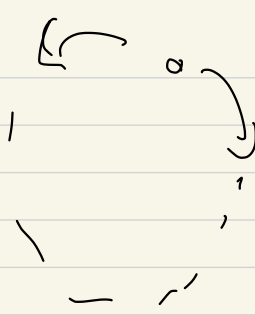
if  $t$  is the Prob<sup>s</sup> of point for  
Gamblers ruin.



Time to reach 0 or n

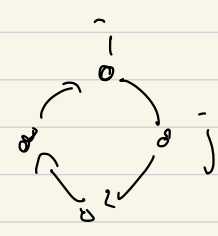
$$t \sim k(n-k)$$

Mixing time or  $\rho = (C_n + C_n^{-1})^{\frac{1}{2}}$



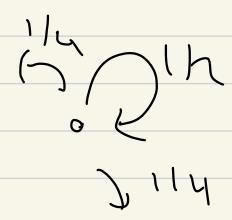
$n$  even

many time =  $\infty$



how

$$P = \frac{1}{4} (C_n + C_n^{-1} + 2I)$$



$$\max_{i,j} |e_i^t P^t - e_j^t P^t| \leq \frac{1}{4} \text{ when}$$

$t$  is the  $\frac{1}{4}$  point of gem run

$\epsilon \rightarrow 0$  mixing time  $\rightarrow \infty$

$$S_{\text{avg}} = \frac{(1-\epsilon)}{2} C_n + \frac{(1-\epsilon)}{2} C_n^{-1} + \epsilon \bar{J}$$

$$\rho = \frac{2}{5} C_n + \frac{2}{5} C_n^{-1} + \frac{1}{5} \bar{J}$$

mixing time?

lazy prob  $\frac{1}{5}$

X, Y!

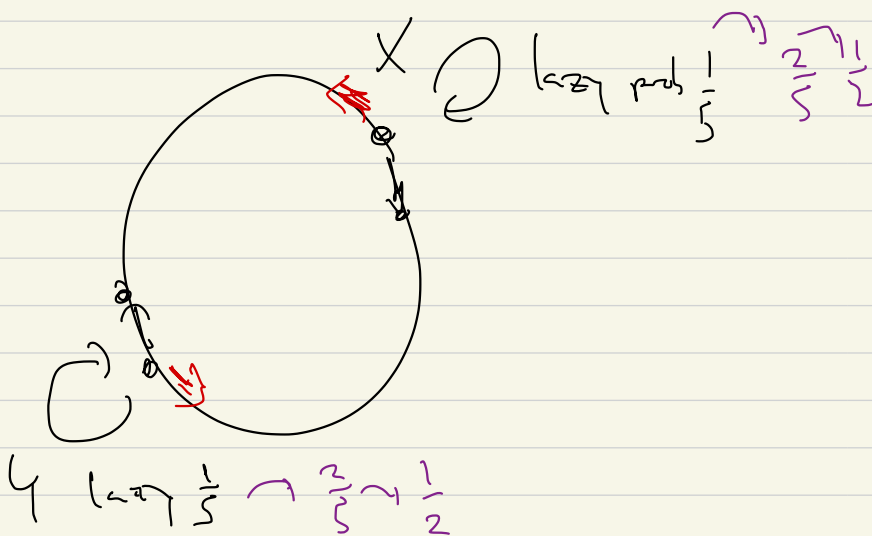
when X is lazy, pr  $\frac{1}{5}$ , Y moves

" Y " " " pr  $\frac{1}{5}$ , X moves

$\frac{3}{5}$   $\frac{2}{5}$  X, Y move in opp dir

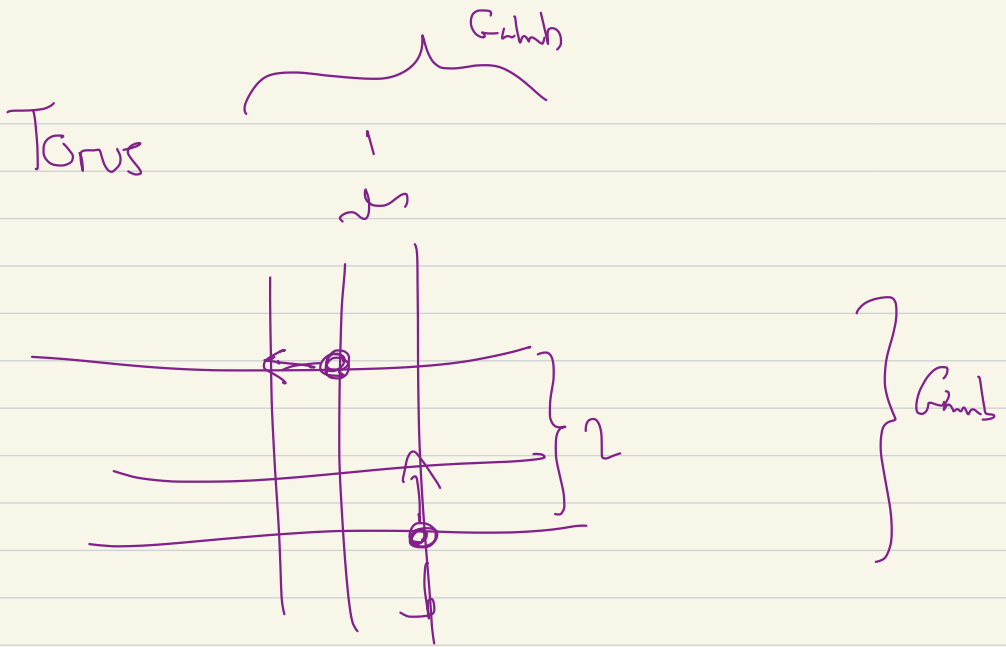
$\frac{2}{5}$  prob! gamblers win move

$\frac{3}{5}$  prob: -- -- stay



after  $t \approx n$  steps of

binomial



Copy!  $X, Y$

pick 1st coord or 2nd coord prob  $\frac{1}{2}$  each

then  
 or  $X$  is lazy prob  $\frac{1}{2}$ ,  $Y$  moves  
 — — —



↓  $A_{11}^{-1}$

↓  $A_{22}^{-1}$



↑  
overlap

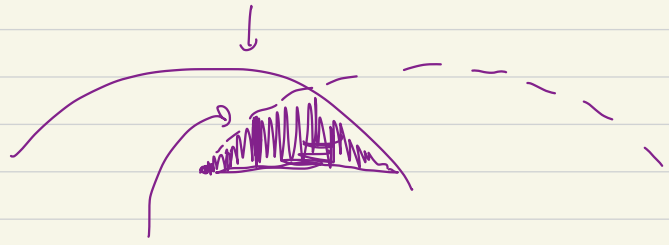
“ shows ”

TV

↔

occupy

prob  $\frac{1}{3}$



prob  $\frac{1}{6}$