

April 6, 2021 CPSC 531F

- Deadline for HW submission:

Sunday, April 25, 11:59 pm

(earlier submissions would be appreciated)

- Perron-Frobenius theorem!

in the context of information

theory, and Markov chains

=

Last class on next Tuesday.

# Information Theory (Shannon):

Say you want encode binary

data,  $\{0,1\}^{n_1}$  into  $\{0,1\}^{n_2}$

data with some constraints.

E.g.  $(d,k)$ -run length constrained

data:

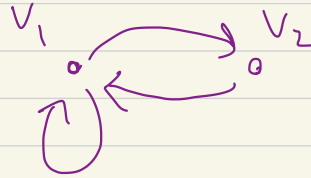
between any two consecutive 1's

there are  $\geq d$  0's

$\leq k$  0's

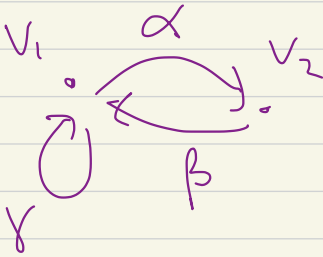
E.g.

Fibonacci graph



$$A_G = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$

this graph can generate strings:



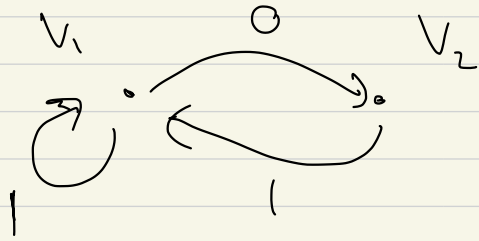
walking from  $v_1$ :

$\gamma\gamma\gamma\alpha\beta\alpha\beta$

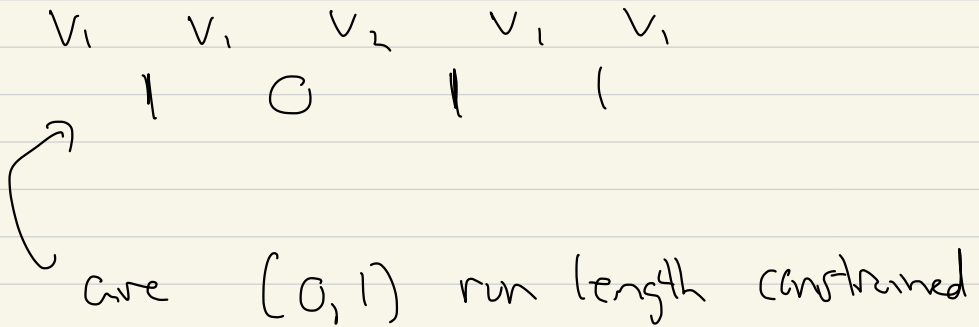
$\gamma\alpha\beta$

$\gamma\gamma\gamma\alpha\beta\gamma\alpha$

Fibonacci  
graph!



gives strings, start walking at  $v_1$



between consecutive 1's have  
either 0, 1 occurrences of 0's



(2,7) graph!

walking at  $v_1$ :  $00100001001-$

Start with  $00$

walking at  $v_3$ : can begin

$v_3$     $v_1$     $v_2$     $v_3$     $v_4$     $v_5$     $v_1$   
|   0   0   0   0   |   ..

# walks of length  $n$  from  $v_1$  (or  $v_2, \dots$ )  $\leq$  # of  $\{0,1\}^n$  strings  $\leq$  # walks of length  $n$  starting from anywhere

$O(\text{sim!})$

#  $\{0,1\}^h$

strings

with

$(2,7)$ -constraint

$$\sim \downarrow \lambda_1(A_G)^h$$

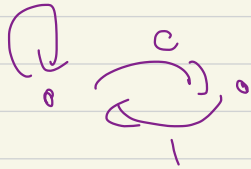
# of  $\{c,1\}^h$

strings

with

$(c,1)$ -constraint

$$\sim \downarrow \lambda_1(A_{\text{Fibonacci}})^h$$



$$A_{\text{Fibonacci}} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$

Based on! Perron-Frobenius theorem:

let  $A \in M_n(\mathbb{R})$  have

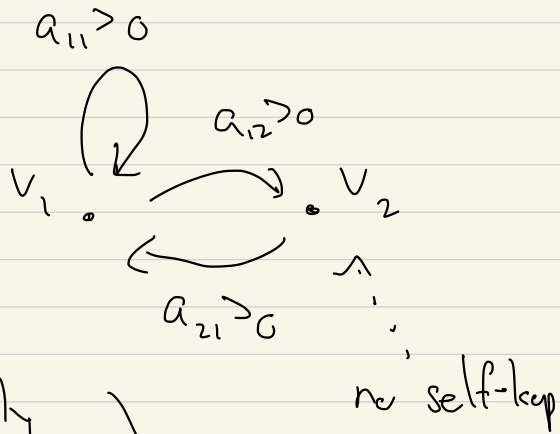
non-neg entries, and let  $A$

be irreducible, i.e. the digraph

associated to  $A$  is strongly

connected

$$\begin{bmatrix} .5 & .3 \\ 700 & 0 \end{bmatrix}$$



(A digraph is strongly connected if for each  $v_1, v_2$  vertices, there's a path from  $v_1$  to  $v_2$ .)

since  $a_{21} = 0$



Example!

$$v_1 \quad v_2 \quad A_G = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$

idea!

$\lambda_{PF} = \lambda_{Perron-Frobenius}$ !

$$\text{Max} \rightarrow \left\{ \lambda \text{ s.t. } \exists \vec{v} \geq 0, \vec{v} \neq \vec{0}, \right. \\ \left. A\vec{v} \geq \lambda\vec{v} \right\}$$

(least upper bound)

(supremum)

$A$  stretches each comp of  $\vec{v}$ , by  $\lambda$

Often find  $\lambda_{PF}$  by "power method"

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix}, A_G \begin{bmatrix} 1 \\ 1 \end{bmatrix} =$$

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

$$A_G^2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} = A_G \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$$

$$A_G^3 \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} = \begin{bmatrix} 5 \\ 3 \end{bmatrix}$$

$$A_G^4 \begin{bmatrix} 1 \\ 1 \end{bmatrix} = A_G \begin{bmatrix} 5 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 5 \\ 3 \end{bmatrix} = \begin{bmatrix} 8 \\ 5 \end{bmatrix}$$

$$A_G^n \begin{bmatrix} 1 \\ 1 \end{bmatrix} : \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 5 \\ 3 \end{bmatrix}, \begin{bmatrix} 8 \\ 5 \end{bmatrix}, \begin{bmatrix} 13 \\ 8 \end{bmatrix}, \dots$$

$n=0 \qquad n=1$

"Power method" to find largest  
eigenvalue of  $A \in M_n(\mathbb{C})$ :

pick at "random"  $v_0 \in \mathbb{C}^n$

$v_0, Av_0, A^2 v_0, \dots$

normalize  
 $\rightarrow$

largest eigenvector

(this often works)

---

$\begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 5 \\ 3 \end{bmatrix}, \begin{bmatrix} 8 \\ 5 \end{bmatrix}, \dots$

$$A \begin{bmatrix} 5 \\ 3 \end{bmatrix} = \begin{bmatrix} 8 \\ 5 \end{bmatrix} \geq \lambda \begin{bmatrix} 5 \\ 3 \end{bmatrix}$$

stretching  $\rightarrow$

$$\lambda = \min \left( \frac{8}{5}, \frac{5}{3} \right)$$

$\uparrow \quad \uparrow$

$$\begin{pmatrix} 5 \\ 3 \end{pmatrix} \rightarrow A_{\text{Fib}}^n \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad \text{get}$$

$$\min \left( - , - \right) \leftarrow$$

vectors of  
consecutive Fib  
numbers

$$F_n \sim c \left( \frac{1+\sqrt{5}}{2} \right)^n$$

$$\rightarrow \left( \frac{1+\sqrt{5}}{2} \right)^n$$

$$\begin{bmatrix} F_{n+1} \\ F_n \end{bmatrix} \stackrel{\text{large } n}{\approx} \begin{bmatrix} F_n \cdot \left(\frac{1+\sqrt{5}}{2}\right) \\ F_n \end{bmatrix}$$

$$A_{\text{Fib}} \begin{bmatrix} \frac{1+\sqrt{5}}{2} \\ 1 \end{bmatrix} = \begin{bmatrix} 1+\sqrt{5} \\ 2 \end{bmatrix} \begin{bmatrix} \frac{1+\sqrt{5}}{2} \\ 1 \end{bmatrix}$$

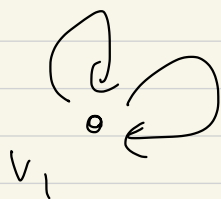
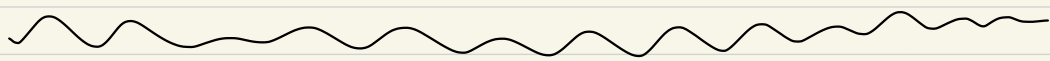
largest eigenvalue  
of  $A_{\text{Fib}}$ .

Define: The information of  $G$  digraph  
capacity

$\log_2(\lambda_{\text{PF}}(G))$  "bits" ( $G$  strongly connected)

Capacity of  $\mathbb{Q}^c$

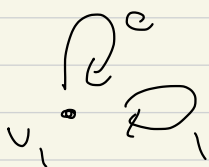
$$:= \log_2 \left( \frac{1+\sqrt{5}}{2} \right) = \dots$$



$$A_G = [2]$$

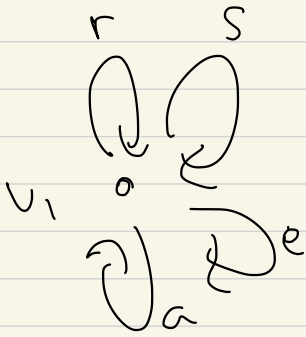
$$\lambda_{\text{PF}} = 2$$

$$\text{Capacity} = \log_2 2 = 1 \quad \text{bit}$$



walks of length  $n$

$\Leftrightarrow$   $\{0, 1\}$  strings  
of length  $n$



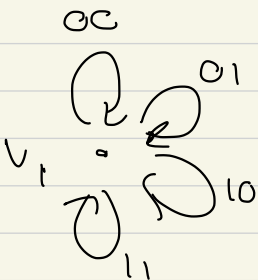
$$A_G = \begin{bmatrix} 4 \\ 4 \end{bmatrix}, \quad |A_{PF} = 4$$

Capacity:  $\log_2 |A_{PF}| = \log_2 4 = 2$

bits

# walks of length  $n \approx 4^n$

rsae rsae rrs ---



produces

# walks of length  $n \approx 4^n = 2^{2n}$

labelling gives

$\{ -1 \}$   
 $\longleftrightarrow$  binary strings  
of length  $2n$

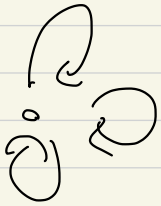
=



$$A_G = [8],$$

$$\text{capacity} = \log_2 8 = 3$$

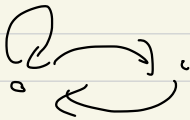
=



$$A_G = [3],$$

$$\text{capacity} \quad \log_2 3 \quad \text{bits}$$

=



$$A_G = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix},$$

$$\log_2 (\lambda_{\text{PF}}(A))$$

is reasonable?



## 2 Tasks!

- if  $A$  is any irreducible matrix with non-negative entries!

$\lambda_{\text{PF}}(A)$  is really an

eigenvalue,  $> 0$  if  $A$  not zero,

# walks length  $n$  in  $A$   $\sim \lambda_{\text{PF}}(A)^n \cdot C$

---

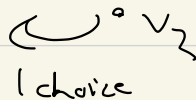
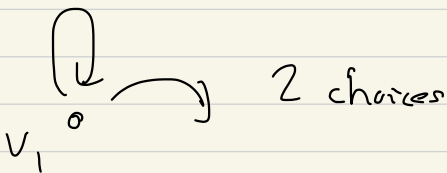
- apply this to convert binary data into constrained data

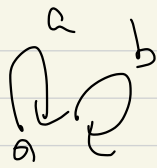
4-minute break

$$\frac{1+\sqrt{5}}{2} \approx 1.618033988\dots$$

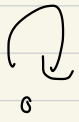
$$\frac{\log \frac{1+\sqrt{5}}{2}}{\log 2} = 0.69424\dots \geq \frac{2}{3}$$

$$\text{Capacity of Fib} = 0.69424\dots \text{ bits}$$

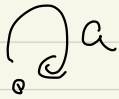




2 choices, capacity  $\log_2 2 = 1$

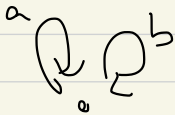


1 choice, capacity  $\log_2 1 = 0$



word aaaaaa

# words length  $n = 1$



aabbabbabbccb...

Fib graph  $\rightarrow > \frac{2}{3}$  bits

3 (Fib graph)  $\rightarrow$  a little more 2 bits..

Data!

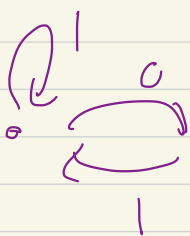
$\{c, 1\}^{2m}$

$2m$  bits of  
arbitrary data

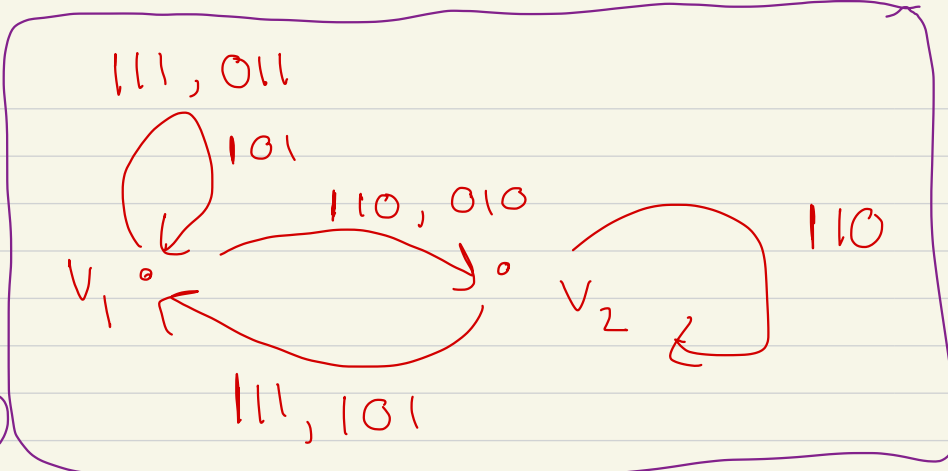
Convert to

$3m$  bits of fib data  
( Since  $\log_2(\lambda_{PF}) > \frac{2}{3}$  )

Walks of length 3 in Fib graph

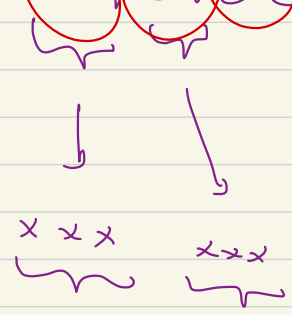


walks of length 3



$A$  walks of length 3 or Fib  $\therefore \left[ A_{Fib} \right]^3$

01010011001

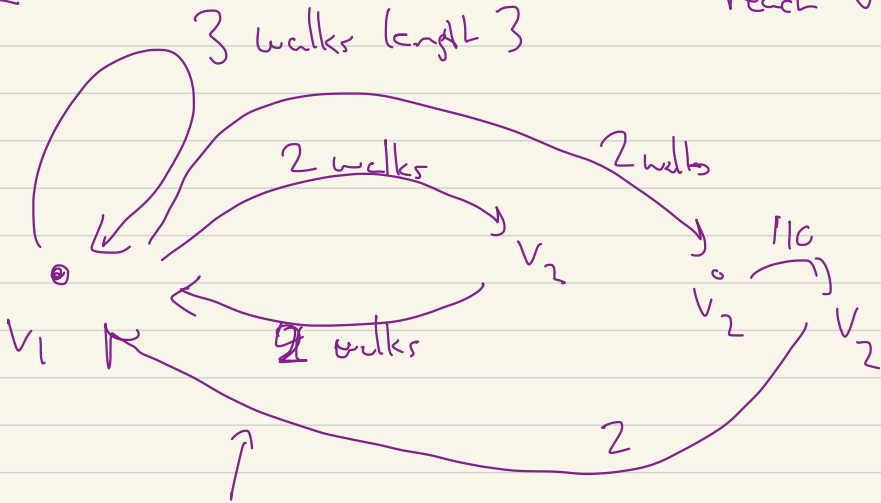


convert arbitrary data

Fib data

ratio of 2 symbols to 3 symbols

[ start at  $V_1$ , end at  $V_1$ , stop first time reach  $V_1$  ]



4 walks length 6

4 walks length 9

3 walks length 3

111

←

2  
11  
00

011

←

01

101

←

10

4 walks length 6

110 111

←

11 00

010 111

←

11 01

110 101

←

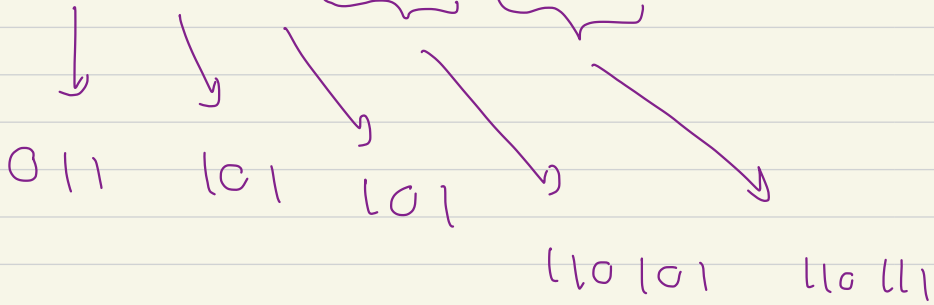
11 10

010 101

←

11 11

0110101110110011 ...



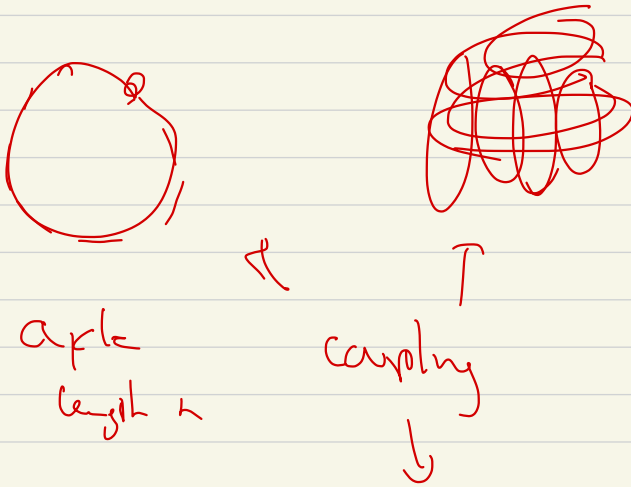
"sliding block" encoding

$$A_{\text{fib}}^3 = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$

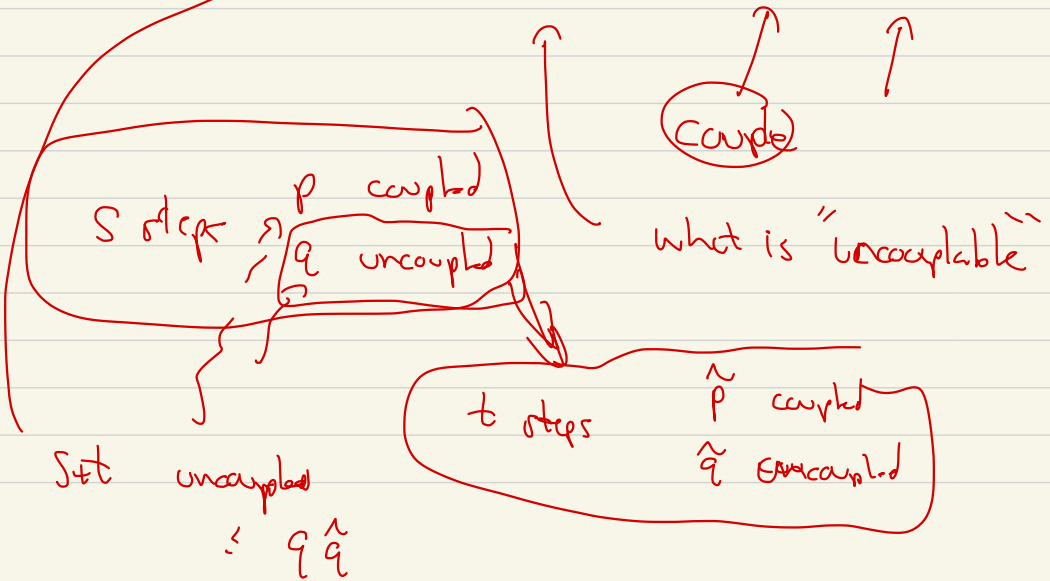
$$\begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 3 & 2 \\ 2 & 1 \end{bmatrix} \leftarrow 5 \text{ choices}$$

$\leftarrow 3 \text{ choices}$

# Class Ends



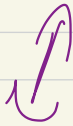
$$\bar{d}(st) \leq \bar{d}(s) \bar{d}(t)$$





$\mathbb{R}^n$ : special eigenvectors)

$$\begin{array}{ccc} | & & | \\ | & & | \\ \text{or } \otimes & & \text{or } \otimes \quad \_ \_ \\ | & & | \\ -1 & & -1 \end{array}$$



$$\chi: (\mathbb{Z}/2\mathbb{Z})^n \rightarrow \mathbb{P}^{\times} \text{ really } \{\pm 1\}$$

$$\min \|A - u_1 v_1^T\|$$

$$A^T A v_1 = \lambda v_1 \quad \lambda \geq 0$$

$$A A^T u_1 = \lambda u_1 \quad \Rightarrow$$

$$\text{tr}(A - u_1 v_1^T)(A - u_1 v_1^T)^T$$

$$= \text{tr}(A A^T) - \lambda$$

$$A - u_1 v_1^T - u_2 v_2^T$$

$$A_1 := A - u_1 v_1^T$$

$$A_1 v_1 = (A - u_1 v_1^T) v_1$$

$$= A v_1 - u_1 (v_1^T v_1)$$

$$= \underbrace{A v_1}_{= 0} - u_1 (v_1^T v_1) = 0$$

$$A_1 \vec{v}_1 = 0$$

$$\vec{u}_1^T A_1 = 0$$

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

$$A^T A \vec{v}_2 = 0 \cdot \vec{v}_2$$

$$A^T \vec{0} = \vec{0}$$

$$A^T A \vec{v}_2$$

Symm  $A^T A$ , eigenvalues

$$0 \leq \lambda_n \leq \dots \leq \lambda_2 \leq \lambda_1$$

$$A^T A \quad \lambda_1$$

$$0 \leq \lambda_n \leq \dots \leq \lambda_3 \leq \lambda_2$$

$$A^T A = \sum_{i=1}^n \lambda_i \vec{v}_i \vec{v}_i^T$$

$$A_1^T A_1 = \begin{pmatrix} \sum & & & \\ & \dots & & \\ & & & \dots \\ & & & & \dots \end{pmatrix}$$

If  $A_1 \neq 0$ ,  $\lambda_2 > 0$

$\vec{v}_1, \dots, \vec{v}_n$  still ON eigenbasis  
 $A A$   $\lambda_1, \lambda_2, \dots, \lambda_n$   
 $A_1^T A_1$   $0, \lambda_2, \dots, \lambda_n$   
 $\nwarrow$   $\uparrow$

$u_2$  picks off largest eigenvalue

---

A :

$\vec{v}_1$  maximizes  $R_A$

$\Rightarrow$

$\vec{v}_2$  "  $R_{A_1}$

$$A_1 = A - \lambda_1 \vec{v}_1 \vec{v}_1^T$$

---

$\lambda_2 = \lambda_3 = \dots$

$$A^T A = \begin{bmatrix} 5 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}$$

$$A_1^T A_1 = \begin{bmatrix} 0 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \quad \vec{v}_2 \in \text{sp}(\vec{e}_2, \vec{e}_3, \vec{e}_4)$$

$$\left( \text{Frob} \left( A_1 - \vec{u}_2 \vec{v}_2^T \right) \right)^2$$

$$\text{Tr} \left( \left( A_1 - \vec{u}_2 \vec{v}_2^T \right) \left( A_1^T - \vec{v}_2 \vec{u}_2^T \right) \right)$$

$$R_{\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}} = \frac{x_1^2 + 2x_2^2 + 3x_3^2}{x_1^2 + x_2^2 + x_3^2}$$

$$\nabla \left( \quad \right) = \nabla \left( \quad \right) \text{ set } (0,0,0)$$



$$\begin{array}{c}
 \hat{A} \\
 \downarrow \\
 \left( \hat{A} - \omega_1 z_1^T - \dots - \omega_{i-1} z_{i-1}^T - \omega_i (z_i + \varepsilon \vec{u})^T \right)^T \\
 \left( - \omega_{i+1} z_{i+1}^T - \dots \right)
 \end{array}$$

$z_1^T + \varepsilon_i \vec{u}_1$   
 $z_i^T + \varepsilon_i$

vary  $z_i \rightsquigarrow (z_i + \varepsilon \vec{u})^T = z_i(\varepsilon)$

$$\vec{u} \perp z_1, \dots, z_{i-1}, z_{i+1}, \dots$$

$$z_1, \dots, z_{i-1}, z_i(\varepsilon), z_{i+1}, \dots \text{ not orth}$$

$$g(\varepsilon) = \left( \hat{A} - \omega_i (z_i + \varepsilon \vec{u})^T \right) \left( \hat{A} - \omega_i (z_i + \varepsilon \vec{u})^T \right)^T$$

↑

the order  $\varepsilon$  term vanishes



$$\text{Tr} \left( \left( \hat{A} - w_i (z_i)^T \right) \cdot u w_i^T \right)$$

+ same = 0

$$\text{Tr} \left( \underbrace{\hat{A} u w_i^T}_{\text{same}} - w_i \underbrace{z_i^T u}_{z_i \cdot u} w_i^T \right)$$

$$\text{Tr}(w_i w_i^T) =$$

$$\text{Tr}(w_i^T w_i) = \overbrace{w_i \cdot w_i}$$

$$w_i^T \hat{A} u = (z_i \cdot u) (w_i \cdot w_i)$$