

# CH5 - P1

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Eigenvalues and eigenvectors of  
a symmetric matrix

Given  $A \in \mathbb{R}^{n \times n}$ , if

$$Av = \lambda v \quad \exists v \in \mathbb{C}^n, v \neq 0$$

(i)  $\lambda \in \mathbb{C}$  is an eigenvalue of  $A$

(ii)  $v \in \mathbb{C}^n$  is an eigenvector corresponding to  $\lambda$ .

First order <sup>(linear)</sup> differential equations  
with constant coefficients

$$(+) \quad \frac{dx}{dt} = Ax$$

$\downarrow$   
 $n \times n$   
constant (independent of  $t$ )

$x: \mathbb{R} \rightarrow \mathbb{C}^n$  (solution sought)

If  $A$  is diagonal,

$$\frac{dx_1}{dt} = a_{11} x_1 \Rightarrow x_1(t) = c_1 e^{a_{11}t}$$

$\vdots$

$$\frac{dx_n}{dt} = a_{nn} x_n \Rightarrow x_n(t) = c_n e^{a_{nn}t}$$

$\exists c_1, \dots, c_n$  (scalars) ①

Now suppose

$$A = V D V^{-1}$$

for some invertible  $V \in \mathbb{C}^{n \times n}$   
diagonal  $D \in \mathbb{C}^{n \times n}$

(+) can be rewritten as

$$\frac{d\underline{y}}{dt} = D \underline{y}$$

$$\text{with } \underline{y} := V^{-1} x.$$

Hence,

$$\underline{y}(t) = \begin{bmatrix} c_1 e^{d_{11}t} \\ \vdots \\ c_n e^{d_{nn}t} \end{bmatrix}$$

$$x(t) = V \underline{y}(t)$$

$$= c_1 e^{d_{11}t} v_1 + \dots + c_n e^{d_{nn}t} v_n$$

Ex

$$\frac{dx(t)}{dt} = \underbrace{\begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}}_A x(t)$$

eigenvalues of  $A$   
 $\lambda_1 = 4$     $\lambda_2 = -2$

corresponding eigenvector  
 $v_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$     $v_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

$$x(t) = c_1 e^{4t} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 e^{-2t} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

for some constants  $c_1, c_2$ .

From here on assume  $A$  is symmetric, that is  $A^T = A$ .

Notation:

$$\mathbb{R}_{\text{sym}}^{n \times n} := \{A \in \mathbb{R}^{n \times n} \mid A^T = A\}$$

Recall for  $A \in \mathbb{R}^{n \times n}$ ,

\* its eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$  are all real,

\* the corresponding eigenvectors  $v_1, v_2, \dots, v_n$  can be chosen so that

$$\{v_1, \dots, v_n\}$$

is orthonormal

Power Iteration

Generates a sequence  $\{q^{(k)}\}$  in  $\mathbb{R}^n$  s.t.

$$q^{(k+1)} := \frac{A q^{(k)}}{\|A q^{(k)}\|}$$

— let us use 2-norm, but other norms are also fine

(for a given  $q^{(0)} \neq 0$ .)

What happens to  $q^{(k)}$  as  $k \rightarrow \infty$ ?

$$q^{(k)} = s_k A^k q^{(0)}$$

$$\text{where } s_k := 1 / \|A^k q^{(0)}\|$$

Eigenvalues in sorted order, that is

$$|\lambda_1| \leq |\lambda_2| \leq \dots \leq |\lambda_{n-1}| \leq |\lambda_n|$$

Expand  $q^{(0)}$  in terms of  $v_1, \dots, v_{n-1}, v_n$

$$q^{(0)} = \alpha_n v_n + \alpha_{n-1} v_{n-1} + \dots + \alpha_1 v_1$$

$$A q^{(0)} = \alpha_n \lambda_n v_n + \alpha_{n-1} \lambda_{n-1} v_{n-1} + \dots + \alpha_1 \lambda_1 v_1$$

$\vdots$

$$A^k q^{(0)} = \alpha_n \lambda_n^k v_n + \alpha_{n-1} \lambda_{n-1}^k v_{n-1} + \dots + \alpha_1 \lambda_1^k v_1$$

Hence,

$$(+++) q^{(k)} = s_k \lambda_n^k \left\{ \alpha_n v_n + \alpha_{n-1} \left( \frac{\lambda_{n-1}}{\lambda_n} \right)^k v_{n-1} + \dots + \alpha_1 \left( \frac{\lambda_1}{\lambda_n} \right)^k v_1 \right\}$$

Assumptions:  $\alpha_n \neq 0$  and  $|\lambda_{n-1}| < |\lambda_n|$

Some observations

$$\textcircled{1} \quad |s_k \lambda_n^k| < \frac{1}{|\alpha_n|} \quad k=1, 2, 3, \dots$$

$$\textcircled{2} \quad \| q^{(k)} - s_k \lambda_n^k \alpha_n v_n \| \rightarrow 0 \quad \text{as } k \rightarrow \infty$$

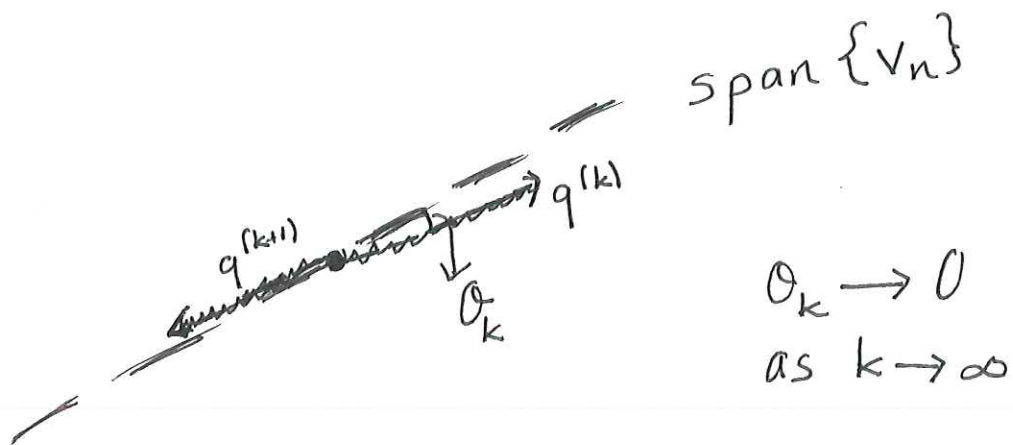
$$\textcircled{3} \quad |s_k \lambda_n^k \alpha_n| \rightarrow 1 \quad \text{as } k \rightarrow \infty$$

## Conclusion (Convergence)

There exists a sequence  $\{c_k\}$  in  $\mathbb{R}$  s.t.

$$(i) \lim_{k \rightarrow \infty} |c_k| = 1, \text{ and}$$

$$(ii) \lim_{k \rightarrow \infty} \|q^{(k)} - c_k v_n\|_2 = 0.$$



## Order of Convergence

It follows from (++) on page (4) that

$$\|q^{(k)} - s_k \lambda_n^k \alpha_n v_n\| = s_k \lambda_n^k \alpha_n \left\{ \frac{\alpha_{n-1}}{\alpha_n} \left( \frac{\lambda_{n-1}}{\lambda_n} \right)^k v_{n-1} + \dots + \frac{\alpha_1}{\alpha_n} \left( \frac{\lambda_1}{\lambda_n} \right)^k v_1 \right\}$$

Additional assumptions:  $\alpha_{n-1} \neq 0$  and  $|\lambda_{n-1}| < |\lambda_n|$

$$\lim_{k \rightarrow \infty} \frac{\|q^{(k+1)} - s_{k+1} \lambda_n^{k+1} \alpha_n v_n\|}{\|q^{(k)} - s_k \lambda_n^k \alpha_n v_n\|} = \lim_{k \rightarrow \infty} \frac{\left\| \frac{\alpha_{n-1}}{\alpha_n} \left( \frac{\lambda_{n-1}}{\lambda_n} \right)^{k+1} v_{n-1} + \dots + \frac{\alpha_1}{\alpha_n} \left( \frac{\lambda_1}{\lambda_n} \right)^{k+1} v_1 \right\|}{\left\| \frac{\alpha_{n-1}}{\alpha_n} \left( \frac{\lambda_{n-1}}{\lambda_n} \right)^k v_{n-1} + \dots + \frac{\alpha_1}{\alpha_n} \left( \frac{\lambda_1}{\lambda_n} \right)^k v_1 \right\|} = \left| \frac{\lambda_{n-1}}{\lambda_n} \right| \quad (5)$$



$E_x$

$$A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix}$$

$$B = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$\lambda_1 = -2 \quad \lambda_2 = 4$$

$$\lambda_1 = 1 \quad \lambda_2 = 3$$

$$v_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad v_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$v_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad v_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Suppose  $q^{(0)} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$\{q^{(k)}\}$  s.t.

$\{p^{(k)}\}$  s.t.

$$q^{(k+1)} = \frac{A q^{(k)}}{\|A q^{(k)}\|}$$

$$p^{(k+1)} = \frac{B p^{(k)}}{\|B p^{(k)}\|}$$

(i)

$\exists \{c_k\}$  s.t.

$\exists \{d_k\}$  s.t.

$$|c_k| \rightarrow 1$$

and

$$\|q^{(k)} - c_k \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}\| \rightarrow 0$$

$$|d_k| \rightarrow 1$$

and

$$\|p^{(k)} - d_k \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}\| \rightarrow 0$$

(ii)

$$\lim_{k \rightarrow \infty} \frac{\|q^{(k+1)} - c_{k+1} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}\|}{\|q^{(k)} - c_k \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}\|} = \left| \frac{-2}{4} \right| = 1/2$$

$$\lim_{k \rightarrow \infty} \frac{\|p^{(k+1)} - d_{k+1} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}\|}{\|p^{(k)} - d_k \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}\|} = \frac{1}{3}$$

# Jacobi's Method

To compute all eigenvalues of  $A \in \mathbb{R}_{sym}^{n \times n}$

Algorithm

Repeat the following for  $k = 0, 1, 2, \dots$  until  $\sum_{\substack{j, l=1 \\ j \neq l}}^n [a_{jl}^{(k)}]^2 < \epsilon$  (letting  $A^{(0)} := A$ )

Find  $p, q \in \{1, 2, \dots, n\}$  but  $p \neq q$  s.t.  
 $|a_{pq}^{(k)}| = \max_{\substack{j, l=1, \dots, n \\ j \neq l}} |a_{jl}^{(k)}|$

$\downarrow$   
prescribed tolerance

$$A^{(k+1)} \leftarrow R^{(pq)}(\theta_k)^T A^{(k)} R^{(pq)}(\theta_k)$$

end

$a_{jl}^{(k)}$  -  $(j, l)$  entry of  $A^{(k)}$

$R^{(pq)}(\theta_k)$  - orthogonal rotation matrix  $(n \times n)$   
s.t.  $a_{pq}^{(k+1)} = 0$ .

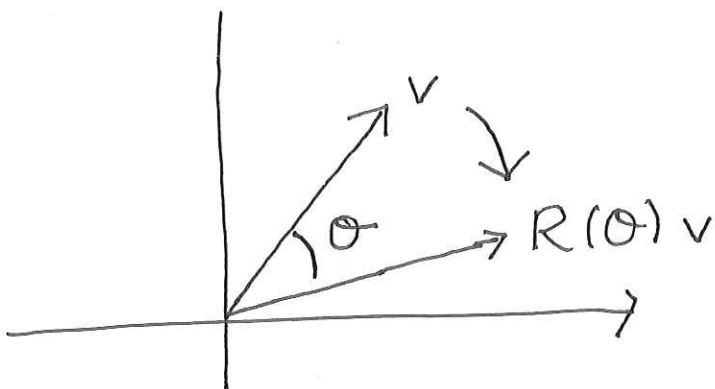
## Remarks

(1)  $A^{(k)}$  is symmetric  $\forall k$ .

(2)  $A, A^{(1)}, \dots, A^{(k)}$   $\forall k$  are similar with the same eigenvalues

# Rotation Matrices

## Rotation in $\mathbb{R}^2$



Rotate  $v$  by  
an angle of  $\theta$  in  
clock-wise direction

$$v \mapsto T_\theta(v)$$

is linear.

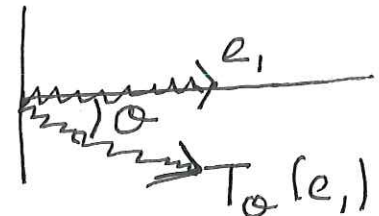
$$T_\theta(v) = R(\theta)v \quad \exists R(\theta) \in \mathbb{R}^{2 \times 2}$$

Due to linearity

$$\begin{aligned} T_\theta(v_1 e_1 + v_2 e_2) &= v_1 T_\theta(e_1) + v_2 T_\theta(e_2) \\ &= \underbrace{\begin{bmatrix} T_\theta(e_1) & T_\theta(e_2) \end{bmatrix}}_{R(\theta)} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \end{aligned}$$

where

$$T_\theta(e_1) = \begin{bmatrix} \cos \theta \\ -\sin \theta \end{bmatrix}$$



$$T_\theta(e_2) = \begin{bmatrix} \sin \theta \\ \cos \theta \end{bmatrix}$$



Hence,

$$R(\theta) = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix}$$

e.g.  $R(\pi/6) = \begin{bmatrix} \sqrt{3}/2 & 1/2 \\ -1/2 & \sqrt{3}/2 \end{bmatrix}$  rotates by  $\pi/6$   
in clock-wise direction

Remark:  $R(\theta)$  is orthogonal, that is

$$R(\theta) R(\theta)^T = R(\theta)^T R(\theta) = I_2$$

(2)



# Planar Rotation in $\mathbb{R}^n$ on the $x_p x_q$ -plane

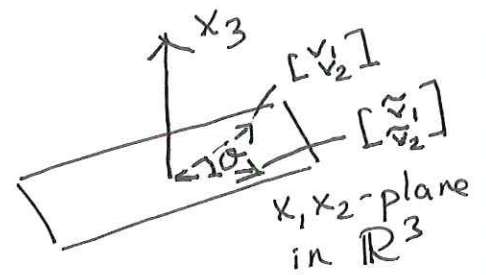
$$R^{(pq)}(\theta) = \begin{bmatrix} \mathbf{I} & & & \\ & c & & s \\ & & \mathbf{I} & \\ & -s & & c \\ & \downarrow & & \downarrow & \mathbf{I} \\ & \text{p}^{\text{th}} & & \text{q}^{\text{th}} & \\ & \text{column} & & \text{column} & \end{bmatrix}$$

$c := \cos(\theta)$   
 $s := \sin(\theta)$

p<sup>th</sup> row  
 q<sup>th</sup> row  
 void entries are 0

$$\tilde{v} := R^{(pq)}(\theta) v$$

$$\begin{cases} \tilde{v}_j = v_j & j = 1, \dots, n, j \neq p, j \neq q \\ \tilde{v}_p = c v_p + s v_q \\ \tilde{v}_q = -s v_p + c v_q \end{cases}$$



## Remark

$R^{(pq)}(\theta)$  is orthogonal, that is

$$\mathbf{I}_n = R^{(pq)}(\theta)^T R^{(pq)}(\theta) = R^{(pq)}(\theta) R^{(pq)}(\theta)^T$$

Determining  $\theta$  s.t.  $(p, q)$  entry of

$$B := R^{(pq)}(\theta)^T \underbrace{A R^{(pq)}(\theta)}_{\tilde{A}}$$

is zero.

p<sup>th</sup> ~~row~~ <sup>column</sup> of  $\tilde{A}$

$$\tilde{a}_{jp} = c a_{jp} - s a_{jq} \quad (j=1, \dots, n)$$

q<sup>th</sup> column of  $\tilde{A}$

$$\tilde{a}_{jq} = s a_{jp} + c a_{jq} \quad (j=1, \dots, n)$$

$p$ th row of  $B$

$$b_{pj} = c \tilde{a}_{pj} - s \tilde{a}_{qj} \quad (j=1, \dots, n)$$

$q$ th row of  $B$

$$b_{qj} = s \tilde{a}_{pj} + c \tilde{a}_{qj} \quad (j=1, \dots, n)$$

Only  $p$ th and  $q$ th rows and columns of  $B$  are different compared with  $A$ .

Need to determine  $c$  and  $s$  (and  $\theta$ ).  
Should exploit  $b_{pq} = 0$ .

$$0 = b_{pq} = c \tilde{a}_{pq} - s \tilde{a}_{qq}$$

$$= c \{s a_{pp} + c a_{pq}\}$$

$$- s \{s a_{qp} + c a_{qq}\}$$

Recall  
 $a_{pq} = a_{qp}$

$$= cs(a_{pp} - a_{qq}) + (c^2 - s^2)a_{pq}$$

$$\implies \frac{2cs}{c^2 - s^2} = \frac{2a_{pq}}{a_{qq} - a_{pp}}$$

$$\left( \frac{2 \cos \theta \sin \theta}{\cos^2 \theta - \sin^2 \theta} = \tan 2\theta \right)$$

$$\implies \theta = \frac{1}{2} \arctan \frac{2a_{pq}}{a_{qq} - a_{pp}}$$

$\downarrow$   
 $\in [-\pi/4, \pi/4]$        $\in [-\pi/2, \pi/2]$

$$c = \cos \theta$$

$$s = \sin \theta$$

if  $a_{pp} = a_{qq}$   
 $\theta = \pi/4$   
or  $\theta = -\pi/4$   
depending  
on sign  
of  $a_{pq}$

## Convergence

The Frobenius norm of  $A \in \mathbb{R}^{n \times n}$

$$\|A\|_F := \sqrt{\sum_{j,k=1}^n a_{jk}^2}$$

is invariant under orthogonal similarity transformations.

### THM

Let  $A \in \mathbb{R}^{n \times n}$  and  $B = Q^T A Q \in \mathbb{R}^{n \times n}$  for some orthogonal matrix  $Q \in \mathbb{R}^{n \times n}$ . Then

$$\|A\|_F = \|B\|_F.$$

### Proof

$$\|B\|_F^2 = \text{Trace}(B^T B)$$

$$= \text{Trace}(Q^T A^T Q \cdot Q^T A Q)$$

$$\left( \begin{array}{l} \text{as} \\ QQ^T = I \end{array} \right) = \text{Trace}(Q^T A^T A Q)$$

$$\left( \begin{array}{l} \text{as} \\ \text{Traces of} \\ \text{2 similar} \\ \text{matrices} \\ \text{are same} \end{array} \right) = \text{Trace}(A^T A) = \|A\|_F^2$$

$$\implies \|A\|_F = \|B\|_F. \quad \square$$

### THM

The sequence  $\{A^{(k)}\}$  of matrices by Jacobi's method is such that

$$\lim_{k \rightarrow \infty} \sum_{\substack{i,j=1 \\ i \neq j}}^n [a_{ij}^{(k)}]^2 = 0.$$

# Proof

Define

$$S(A^{(k)}) := \sum_{i,j=1}^n [a_{ij}^{(k)}]^2 = \|A^{(k)}\|_F^2$$

$$D(A^{(k)}) := \sum_{i=1}^n [a_{ii}^{(k)}]^2$$

$$L(A^{(k)}) := \sum_{\substack{i,j=1 \\ i \neq j}}^n [a_{ij}^{(k)}]^2$$

so that  $S(A^{(k)}) = D(A^{(k)}) + L(A^{(k)})$

as well as  $S(A^{(k)}) = S(A^{(k+1)})$ .

Recall

$$\begin{bmatrix} a_{pp}^{(k+1)} & a_{pq}^{(k+1)} \\ a_{qp}^{(k+1)} & a_{qq}^{(k+1)} \end{bmatrix} = \begin{bmatrix} c & s \\ -s & c \end{bmatrix}^T \begin{bmatrix} a_{pp}^{(k)} & a_{pq}^{(k)} \\ a_{qp}^{(k)} & a_{qq}^{(k)} \end{bmatrix} \begin{bmatrix} c & s \\ -s & c \end{bmatrix}$$

Hence, by previous thm

$$\begin{aligned} [a_{pp}^{(k)}]^2 + 2[a_{pq}^{(k)}]^2 + [a_{qq}^{(k)}]^2 &= [a_{pp}^{(k+1)}]^2 + 2[a_{pq}^{(k+1)}]^2 + [a_{qq}^{(k+1)}]^2 \\ &= [a_{pp}^{(k+1)}]^2 + [a_{qq}^{(k+1)}]^2. \end{aligned}$$

As all diagonal entries of  $A^{(k)}$  and  $A^{(k+1)}$  are the same, except  $(p,p)$  and  $(q,q)$  entry,

$$\boxed{\text{since } S(A^{(k+1)}) = S(A^{(k)})} \implies D(A^{(k+1)}) = D(A^{(k)}) + 2[a_{pq}^{(k)}]^2$$

$$\implies L(A^{(k+1)}) = L(A^{(k)}) - 2[a_{pq}^{(k)}]^2$$

Now  $L(A^{(k)}) \leq (n^2 - n) [a_{pq}^{(k)}]^2 \implies -2[a_{pq}^{(k)}]^2 \leq \frac{-2}{n^2 - n} L(A^{(k)})$ ,  
so

$$L(A^{(k+1)}) \leq \underbrace{\left(1 - \frac{2}{n^2 - n}\right)}_{\in [0, 1)} L(A^{(k)})$$

⑥

implying

$$L(A^{(k)}) \leq \left(1 - \frac{2}{n^2 - n}\right)^k L(A)$$

so that  $\lim_{k \rightarrow \infty} L(A^{(k)}) = 0$  as desired.  $\square$

## Summary

- ①  $A, A^{(k)} \forall k$  have the same eigenvalues.
- ②  $A^{(k)}$  becomes diagonal as  $k \rightarrow \infty$ .
- ③ diagonal entries of  $A^{(k)}$  become eigenvalues of  $A$  as  $k \rightarrow \infty$ .