

MATH 312  
LECTURE 19

: The spectral theorem.

Continuation ODEs and diff. eq.

• Example: ODE with complex eigenvalues.

Solve  $\vec{x}'(t) = \begin{bmatrix} 2 & -3 \\ 3 & 2 \end{bmatrix} \vec{x}(t)$ , for general initial data  $\vec{x}_0 = \begin{bmatrix} x_0 \\ y_0 \end{bmatrix}$ .

Sol:

$\vec{x}(t) = e^{At} \vec{x}_0$ .

\*Remark: You might prefer to use the method described in the 3y book

$\det(A - \lambda I) = 0 \rightarrow \lambda_1 = 2 + 3i$   
 $\lambda_2 = 2 - 3i$ .

$\lambda_1 = 2 + 3i \rightarrow \vec{u}_1 = \begin{bmatrix} i \\ 1 \end{bmatrix}$ ,  $\lambda_2 = 2 - 3i \rightarrow \vec{u}_2 = \begin{bmatrix} -i \\ 1 \end{bmatrix}$

$e^{At} = \frac{1}{2} \begin{bmatrix} i & i \\ 1 & -1 \end{bmatrix} \begin{bmatrix} e^{2t} e^{3it} & 0 \\ 0 & e^{2t} e^{-3it} \end{bmatrix} \begin{bmatrix} -i & 1 \\ -i & -1 \end{bmatrix} \frac{1}{2} =$

$= \frac{1}{2} \begin{bmatrix} e^{2t} e^{3it} + e^{2t} e^{-3it} & i e^{2t} e^{3it} - i e^{2t} e^{-3it} \\ -i e^{2t} e^{3it} + i e^{2t} e^{-3it} & e^{2t} e^{3it} + e^{2t} e^{-3it} \end{bmatrix} \rightarrow \vec{x}(t) = e^{At} \vec{x}_0 \rightarrow$

$\Rightarrow \vec{x}(t) = \begin{bmatrix} \frac{1}{2} x_0 e^{2t} (\cos(3t) + i \sin(3t)) + \cos(3t) - i \sin(3t) + \frac{1}{2} y_0 e^{2t} (i \cos(3t) - i \sin(3t) - i \cos(3t) - \sin(3t)) \\ \frac{1}{2} x_0 e^{2t} (-i \cos(3t) + \sin(3t) + i \cos(3t) + \sin(3t)) + \frac{1}{2} y_0 e^{2t} (\cos(3t) + i \sin(3t) + \cos(3t) + i \sin(3t)) \\ \vdots \end{bmatrix}$

## VI The Spectral Theorem.

• Theorem: The Spectral Theorem  
(or Principal Axis Theorem)

If  $M$  is a real symmetric  $n \times n$  matrix, then it has real eigenvalues and the eigenvectors can be chosen to be orthonormal. That is,

$M$  can always be diagonalised as

$$M = Q D Q^T, \text{ where}$$

$D \equiv$  diagonal with real eigenvalues.

$Q \equiv$  formed by  $n$  orthonormal eigenvectors.

Remarks:

1) Eigenvectors corresponding to eigenvalues go in the same order:

$$\begin{array}{l} \lambda_1 \leftrightarrow \vec{v}_1 \\ \lambda_2 \leftrightarrow \vec{v}_2 \\ \vdots \end{array} \rightsquigarrow Q = [\vec{v}_1 | \vec{v}_2 | \dots] \quad D = \begin{bmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \ddots \end{bmatrix}$$

⌈ Note that if  $M = Q D Q^T$ , then  $M$  is symmetric:  $(Q D Q^T)^T = Q D Q^T$ .

So  $M = Q D Q^T \iff M$  is symmetric

⌋

2) If the eigenvalues are distinct, then the eigenvectors are automatically perpendicular to each other.

↳ We just need to be careful to make them unit size.  
(i.e., orthonormal).

3) If an eigenvalue is repeated (let's say twice), then the nullspace of  $A - \lambda I$  will have dimension 2 (or more if repeated more times...).

↳ In this case we need to obtain an orthonormal basis of that nullspace (the "eigenspace").

↳ Obtain a basis and use Gram-Schmidt!  
(as usual)

4)  $Q$  has orthonormal columns and it is square. Thus, it automatically has ~~orthogonal~~ orthonormal rows too. This is called orthogonal matrix. [*(\*)* proof = -163]

↳  $Q$  orthogonal  $\Leftrightarrow Q^T Q = Q Q^T = I$  (i.e.,  $Q^T = Q^{-1}$ ).

$Q$  orthonormal columns  $\Leftrightarrow Q^T Q = I$

$Q$  orthonormal rows  $\Leftrightarrow Q Q^T = I$ .

$$\begin{aligned}
 5) \quad M &= Q \Lambda Q^T = \left[ \vec{q}_1 | \dots | \vec{q}_n \right] \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix} \begin{bmatrix} \vec{q}_1^T \\ \vdots \\ \vec{q}_n^T \end{bmatrix} = \\
 &= \left[ \lambda_1 \vec{q}_1 | \dots | \lambda_n \vec{q}_n \right] \begin{bmatrix} \vec{q}_1 \\ \vdots \\ \vec{q}_n \end{bmatrix} = \\
 &= \left\| \lambda_1 \underbrace{\vec{q}_1 \vec{q}_1^T} + \dots + \lambda_n \underbrace{\vec{q}_n \vec{q}_n^T} \right\| = M \quad \left( \begin{array}{l} \text{nice} \\ \text{decomposition} \end{array} \right)
 \end{aligned}$$

$\swarrow \quad \searrow$   
 projection matrices onto the (unitary) axes  
 given by the (orthonormal) eigenvectors.

(\*) Proof of:  $Q$  square with orthonormal columns  $\Rightarrow Q$  has orthonormal rows (and thus,  $Q$  is "orthogonal")

$\hookrightarrow Q^T Q = I$ . Then, we know that  $Q^{-1}$  exists since

$$\det(Q^T Q) = \det(Q^T) \det(Q) = (\det(Q))^2 = \det(I) = 1 \Rightarrow$$

$\Rightarrow \det(Q) = \pm 1 \neq 0$ . (We needed determinants to prove this!)

Therefore, multiplying by  $Q$  we obtain

$$Q Q^T Q = Q \Rightarrow Q Q^T Q Q^{-1} = Q Q^{-1} \Rightarrow Q Q^T = I$$

$\downarrow$   
orthonormal rows

Before proving the theorem, we need to <sup>remember</sup> ~~mind~~ some things about complex numbers:

$$z \in \mathbb{C} \rightarrow z = a + bi, a, b \in \mathbb{R}, i^2 = -1.$$

$$\bullet z_1 + z_2 = (a + bi) + (c + di) = (a + c) + (b + d)i.$$

↑  
def.

$$\bullet z_1 z_2 = (a + bi)(c + di) = (ac - bd) + i(ad + bc).$$

↑  
def.

$$\bullet \bar{z}_1 = a - bi \text{ (conjugation)}$$

↑  
def.

Some more things we need:

$$1) z + \bar{z} = a + bi + a - bi = 2a \in \mathbb{R} \text{ purely real.}$$

$$2) z - \bar{z} = 2ib \text{ purely imaginary}$$

$$3) z = \bar{z} \Leftrightarrow z \in \mathbb{R}.$$

$$4) z\bar{z} = a^2 + b^2 = |z|^2 \text{ (modulus squared)}$$

$$5) \vec{z} = \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} \rightarrow \|\vec{z}\|^2 = \vec{z}^T \vec{z} = \begin{bmatrix} \bar{z}_1 & \bar{z}_2 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} = |z_1|^2 + |z_2|^2.$$

$$6) \overline{z_1 z_2} = \bar{z}_1 \bar{z}_2$$

Proof of Spectral theorem: Let's split it into two parts.

1)  $M$  real and symmetric  $\Rightarrow$  real eigenvalues.

Let  $\lambda, \vec{a}$  be an eigenvalue and its eigenvector (both maybe) complex

Then (we want to prove  $\lambda$  is real, i.e.,  $\lambda = \bar{\lambda}$ )

$$\textcircled{1} M\vec{a} = \lambda\vec{a}$$

Therefore, taking conjugates,  $\overline{M\vec{a}} = \overline{\lambda\vec{a}} \Rightarrow M\bar{\vec{a}} = \bar{\lambda}\bar{\vec{a}} \Rightarrow$   
 $\Rightarrow M\bar{\vec{a}} = \bar{\lambda}\bar{\vec{a}} \quad \uparrow$   
 $M \text{ is real.}$

$$\textcircled{2} M\bar{\vec{a}} = \bar{\lambda}\bar{\vec{a}}$$

$$\rightarrow \text{Multiply } \textcircled{1} \text{ by } \bar{\vec{a}}^T: \bar{\vec{a}}^T M \vec{a} = \bar{\vec{a}}^T \lambda \vec{a} = \lambda \bar{\vec{a}}^T \vec{a} = \lambda \|\vec{a}\|^2$$

$$\rightarrow \text{Multiply } \textcircled{2} \text{ by } \vec{a}^T: \vec{a}^T M \bar{\vec{a}} = \vec{a}^T \bar{\lambda} \bar{\vec{a}} = \bar{\lambda} \vec{a}^T \bar{\vec{a}} = \bar{\lambda} \|\vec{a}\|^2 \quad (*)$$

These are just numbers! (not vectors or matrices)

$$\left[ \begin{array}{c} \bar{\vec{a}}^T \\ \vec{a}^T \end{array} \right] \left[ \begin{array}{c} M \\ M \end{array} \right] \left[ \begin{array}{c} \vec{a} \\ \bar{\vec{a}} \end{array} \right] = \left[ \begin{array}{c} \lambda \\ \bar{\lambda} \end{array} \right] \left[ \begin{array}{c} \|\vec{a}\|^2 \\ \|\vec{a}\|^2 \end{array} \right] = \text{number}$$

Therefore,

$$\bar{\vec{a}}^T M \vec{a} = (\bar{\vec{a}}^T M \vec{a})^T = \vec{a}^T M^T \bar{\vec{a}} = \vec{a}^T M \bar{\vec{a}}$$

$\uparrow \quad \uparrow$   
 same number!  $M$  is symmetric

Thus we have found that

$\bar{u}^T M u = u^T M \bar{u}$  and we have in (\*) that

$$\left. \begin{aligned} \bar{u}^T M u &= \lambda \|u\|^2 \\ u^T M \bar{u} &= \bar{\lambda} \|u\|^2 \end{aligned} \right\}$$

so we conclude that

$$\lambda \|u\|^2 = \bar{\lambda} \|u\|^2 \Rightarrow (\lambda - \bar{\lambda}) \|u\|^2 \Rightarrow \lambda - \bar{\lambda} = 0 \Rightarrow \lambda = \bar{\lambda}$$

$\uparrow$   
( $u \neq \vec{0}$  by  
definition of eigenvector)

$\lambda$  is real)

2)  $M$  real, symmetric  $\Rightarrow$  orthogonal eigenvectors.

⌈ We only prove it for distinct eigenvalues.

But it is always true: if we have ~~an~~ an eigenvalue repeated, for symmetric matrices we always obtain a full set of eigenvectors  $\xrightarrow{\text{so}}$  orthogonal set  
Gram-Schmidt ||

Let  $\lambda_1 \neq \lambda_2$  eigenvalues with corresponding eigenvectors  $\vec{u}_1, \vec{u}_2$ .

[ Remark: We know from 1) that  $\lambda_1$  and  $\lambda_2$  are real (and thus  $\vec{u}_1, \vec{u}_2$ , being solutions of  $(A - \lambda_i I)\vec{u}_i = \vec{0} \dots$ , are also real ). ]

$\rightarrow$  We want to show that  $\vec{u}_1 \cdot \vec{u}_2 = \vec{u}_1^T \vec{u}_2 = \vec{u}_2^T \vec{u}_1 = 0$  :

$$M \vec{u}_1 = \lambda_1 \vec{u}_1 \rightarrow \vec{u}_2^T M \vec{u}_1 = \lambda_1 \vec{u}_2^T \vec{u}_1$$

$$M \vec{u}_2 = \lambda_2 \vec{u}_2 \rightarrow \vec{u}_1^T M \vec{u}_2 = \lambda_2 \vec{u}_1^T \vec{u}_2$$

$$\text{but } \vec{u}_2^T M \vec{u}_1 = (\vec{u}_2^T M \vec{u}_1)^T = \vec{u}_1^T M^T \vec{u}_2 = \vec{u}_1^T M \vec{u}_2$$

$\uparrow$   
same  
number

$\uparrow$   
 $M$  symmetric

$$\Rightarrow \lambda_1 \vec{u}_2^T \vec{u}_1 = \lambda_2 \vec{u}_1^T \vec{u}_2 \Rightarrow (\lambda_1 - \lambda_2) \vec{u}_2^T \vec{u}_1 = 0 \Rightarrow \vec{u}_2^T \vec{u}_1 = 0$$

$\uparrow$   
 $\lambda_1 \neq \lambda_2$

-H1-