
EECS 16B

Spring 2022

Lecture 18

3/17/2022 

LECTURE 18 : - recap of upper triangularization - Spectral Theorem

Last time: any square matrix can be upper triangularized (proven for real matrices with real eigenvalues; true for complex eigenvalues also).

$$U^T A U = T \quad \begin{array}{l} U: \text{orthogonal} \\ T: \text{upper triangular} \end{array}$$

Note: 1) A and T have the same eigenvalues:

if $(\lambda, \vec{v})$ eval/evector for T , then $(\lambda, U\vec{v})$
is " " " for A

$$A(U\vec{v}) = \underbrace{(U^T U)^T}_{=I} (U\vec{v}) = U T \vec{v} = U \lambda \vec{v} = \lambda (U\vec{v})$$

multiply from left by U , right by U^T : $A = U T U^T$

thus, $A(U\vec{v}) = \lambda(U\vec{v})$ i.e. λ is eval for A also.

2) diagonal of entries of T are its evals
(because upper triangular - shown last time)

From (1), (2): once matrix A is upper triangularized its evals appear in the diagonal entries of T :

$$U^T A U = \begin{bmatrix} \lambda_1 & * & \cdots & * \\ 0 & \ddots & & \vdots \\ \vdots & & \ddots & * \\ 0 & \cdots & 0 & \lambda_n \end{bmatrix}$$

4) Induction proof from last lecture can be turned into a recursive algorithm (Algorithm 10 in Note 15):

Define a function:

Given $A \in \mathbb{R}^{(k+1) \times (k+1)}$ with real evals,

- pick eval/evector pair $(\lambda_1, \vec{q}_1)$ such that $\|\vec{q}_1\|=1$
- complete $\vec{q}_1$ to an orthonormal basis for $\mathbb{R}^{k+1}$: $\{\vec{q}_1, \vec{q}_2, \dots, \vec{q}_{k+1}\}$, e.g., by Gram Schmidt.

Define $Q = [\vec{q}_1 \ \vec{q}_2 \ \dots \ \vec{q}_{k+1}]$. Then

$$Q^T A Q = \begin{bmatrix} \lambda_1 & \star \\ \vec{0} & A_0 \end{bmatrix}.$$

- Return Q and A_0 .

Given $n \times n$ matrix A that we want to upper-triangularize:

$(Q, A_0) = \text{function_above}(A)$

$U = Q$

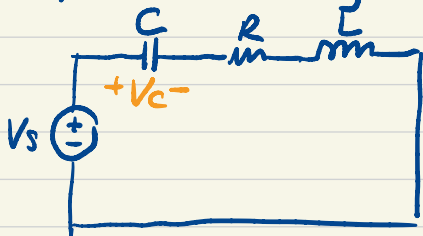
while $\text{size}(A_0) > 1$

$(Q, A_0) = \text{function_above}(A_0)$

$U = U * \begin{bmatrix} 1 & 0 \\ 0 & Q \end{bmatrix}$

end;

Example: (Critically damped) RLC circuit



HW5:

$$x_1(t) = V_c(t)$$

$$x_2(t) = \frac{d}{dt} V_c(t)$$

$$\lambda_{1,2} = \frac{-R}{2L} \pm \frac{1}{2} \sqrt{\frac{R^2}{L^2} - \frac{4}{LC}} \Leftarrow A = \begin{bmatrix} 0 & 1 \\ -\frac{1}{LC} & -\frac{R}{L} \end{bmatrix}$$

Suppose $\frac{R^2}{L^2} = \frac{4}{LC} \Rightarrow \lambda_1 = \lambda_2 = -\frac{R}{2L}$ (HW5 assumed $\lambda_1 \neq \lambda_2$)

$$\hookrightarrow -\frac{1}{LC} = -\frac{R^2}{4L^2}$$

$$\Downarrow A = \begin{bmatrix} 0 & 1 \\ -\frac{R^2}{4L^2} & -\frac{R}{L} \end{bmatrix}$$

Eigenvectors: $\lambda_{1,2} I - A = \begin{bmatrix} -\frac{R}{2L} & \\ & -\frac{R}{2L} \end{bmatrix} + \begin{bmatrix} 0 & -1 \\ \frac{R^2}{4L^2} & \frac{R}{L} \end{bmatrix}$

$$= \begin{bmatrix} -\frac{R}{2L} & -1 \\ \frac{R^2}{4L^2} & \frac{R}{2L} \end{bmatrix}$$

null space?

$$\vec{v} = \begin{bmatrix} 1 \\ -\frac{R}{2L} \end{bmatrix} \cdot \alpha \quad \alpha \neq 0$$

Can't find two lin. ind. eigenvectors.
 $\Rightarrow$ not diagonalizable.

Can nevertheless upper triangularize A :

$$U^T A U = \begin{bmatrix} \lambda & * \\ 0 & \lambda \end{bmatrix}, \quad \lambda = \frac{-R}{2L}$$

for some orthogonal U you will find in HW9.

Sol'n of circuit diff. eq's ($V_S = 0$ for simplicity):

$$\frac{d}{dt} \vec{x}(t) = A \vec{x}(t) \xrightarrow{\vec{y} = U^T \vec{x}} \frac{d}{dt} \vec{y}(t) = U^T A \underbrace{\vec{x}(t)}_{U \vec{y}(t)} \\ = U^T A U \vec{y}(t)$$

$$\frac{d}{dt} y_1(t) = \lambda y_1(t) + * y_2(t)$$

$$\frac{d}{dt} y_2(t) = \lambda y_2(t)$$

$$\rightarrow y_2(t) = e^{\lambda t} y_2(0)$$

$$\frac{d}{dt} y_1(t) = \lambda y_1(t) + * e^{\lambda t} y_2(0)$$

treat as input

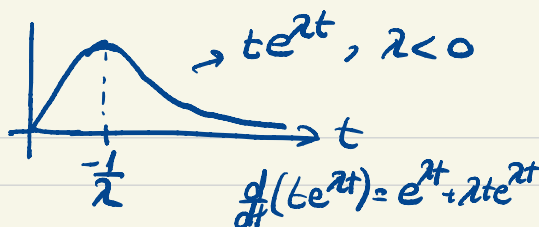
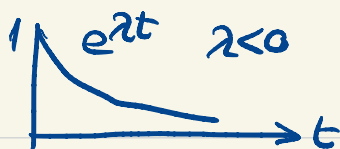
$$y_1(t) = e^{\lambda t} y_1(0) + \int_0^t \underbrace{e^{\lambda(t-\tau)} * e^{\lambda \tau} y_2(0)}_{= e^{\lambda t} * y_2(0)} d\tau$$

$$= e^{\lambda t} * y_2(0) \underbrace{\int_0^t d\tau}_t$$

$$y_1(t) = e^{\lambda t} y_1(0) + t e^{\lambda t} * y_2(0)$$

$$y_2(t) = e^{\lambda t} y_2(0)$$

$\rightarrow$ new! signature of repeated eigenvalue



Similar $t^2 e^{\lambda t}$ would pop up in the solution to:

$$\frac{d}{dt} \vec{y}(t) = \begin{bmatrix} \lambda & * & * \\ 0 & \lambda & * \\ 0 & 0 & \lambda \end{bmatrix} \vec{y}(t)$$

$U^T A U = T \Rightarrow A = U T U^T$ where U : orthogonal
 T : upper triangular
 $\hookrightarrow$ called "Schur decomposition"

Spectral Theorem: Motivation:

- For a diagonalizable matrix A we can find V s.t.

$$V^{-1} A V = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$$

V : not necessarily orthogonal

- If we instead upper triangularize we find **orthogonal** U such that

$$U^{-1} A U = U^T A U = \begin{bmatrix} \lambda_1 & * & * \\ 0 & \ddots & * \\ 0 & & \lambda_n \end{bmatrix}$$

- For symmetric matrices ($A = A^T$) we get both:
 Diagonalizable with an **orthogonal** V :

$$V^{-1} A V = V^T A V = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \ddots & & \\ 0 & & 0 & \lambda_n \end{bmatrix}$$

Theorem (Spectral Theorem):

Let $A \in \mathbb{R}^{n \times n}$ be symmetric. Then:

- (i) The eigenvalues of A are real.
- (ii) A diagonalizable.
- (iii) Eigenvectors of A are pairwise orthogonal (therefore, if we choose them to have length 1 then they constitute an orthonormal matrix:

$$V = [\vec{v}_1 \dots \vec{v}_n] \text{ orthogonal matrix})$$

Proof:

- (i) Let $(\lambda, \vec{v})$ be eigenvalue/eigenvector pair.

$$A\vec{v} = \lambda\vec{v} \quad \dots (1)$$

$$\lambda = a + jb \quad (\text{show } b=0, \text{ i.e., } \bar{\lambda} = a - jb = a = \lambda)$$

Take complex conjugates of both sides of (1):

$$\overline{(A\vec{v})} = \overline{(\lambda\vec{v})}$$

$$= \overline{A}\vec{\bar{v}} = \bar{\lambda}\vec{\bar{v}}$$

$$= A \text{ because } A \text{ real}$$

$$A\vec{\bar{v}} = \bar{\lambda}\vec{\bar{v}}$$

Transpose: $\vec{\bar{v}}^T A^T = \vec{\bar{v}}^T \bar{\lambda}$ $\rightarrow = A$ by symmetry
multiply from right by $\vec{v}$:

$$\vec{\bar{v}}^T A \vec{v} = \bar{\lambda}(\vec{\bar{v}}^T \vec{v}) \quad \dots (2)$$

Multiply (1) from left by $\bar{v}^T$:

$$\bar{v}^T A v = \lambda (\bar{v}^T v) \quad \text{--- (3)}$$

LHS of (2), (3) are the same so RHS must also be same:

$$\bar{\lambda} (\bar{v}^T v) = \lambda (\bar{v}^T v)$$

$$\Rightarrow \underline{\bar{\lambda} = \lambda}$$

$$\Rightarrow \lambda \text{ is real.}$$

$$\begin{aligned} [\bar{v}_1 \dots \bar{v}_n] \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} &= \bar{v}_1 v_1 + \dots + \bar{v}_n v_n \\ &= |v_1|^2 + \dots + |v_n|^2 \\ &\neq 0 \text{ bc } v \text{ is vector} \end{aligned}$$

Evector v also real:

$$(A - \lambda I) v = 0$$

$\downarrow$
real
 $\underbrace{\hspace{1cm}}_{\text{real}}$

(ii) Apply Schur decomposition:

$$\begin{aligned} U^T A U &= T \\ T^T &= U^T A^T (U^T)^T \\ &= U^T A U = T \end{aligned}$$

$T = T^T$ and upper triangular $\Rightarrow T$ diagonal

$$U^T A U = \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$$