

§ 6.4 Hermitian Matrix (Cont.)

Thm If the eigenvalues of a Hermitian matrix A are distinct, then \exists a unitary matrix U that diagonalizes A .

$$\text{Ex: } A = \begin{bmatrix} 2 & 1-i \\ 1+i & 1 \end{bmatrix}$$

$$\lambda_1 = 0, \quad \lambda_2 = 3$$

$$\vec{x}_1 = \begin{bmatrix} -1 \\ 1+i \end{bmatrix}, \quad \vec{x}_2 = \begin{bmatrix} 1-i \\ 1 \end{bmatrix}$$

$$\text{Let } \vec{u}_1 = \frac{1}{\sqrt{2}} \vec{x}_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 1+i \end{pmatrix}$$

$$\text{Since } \|\vec{x}_1\| = \sqrt{\vec{x}_1^H \vec{x}_1} = \sqrt{(-1, 1+i) \begin{pmatrix} -1 \\ 1+i \end{pmatrix}} = \sqrt{1+(-1)(1+i)^2} = \sqrt{3}$$

$$\vec{u}_2 = \frac{1}{\sqrt{3}} \vec{x}_2 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1-i \\ 1 \end{pmatrix}$$

$$\|\vec{x}_2\| = \sqrt{(1-i, 1) \begin{pmatrix} 1-i \\ 1 \end{pmatrix}} = \sqrt{(1-i)(1-i)} = \sqrt{3}$$

The unitary matrix U :

$$U = \begin{bmatrix} \vec{u}_1 & \vec{u}_2 \end{bmatrix} = \frac{1}{\sqrt{3}} \begin{bmatrix} -1 & 1-i \\ 1+i & 1 \end{bmatrix}$$

$$D = \begin{bmatrix} 0 & 0 \\ 0 & 3 \end{bmatrix} \quad \text{or} \quad D = U^H A U = \dots$$

$$\text{So } A = \frac{1}{\sqrt{3}} \begin{bmatrix} -1 & 1-i \\ 1+i & 1 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} -1 & 1-i \\ 1+i & 1 \end{bmatrix}$$

Thm (Schur's Thm)

For each non matrix A , there exists a unitary matrix U s.t.

$T = U^H A U$ is upper triangular.

(proof in the textbook) 數量归納法 + 極值法

Remark: The factorization $A = U T U^H$ is called the Schur decomposition of A .

In the case A is Hermitian, the matrix T will be diagonal.

Thm (Spectral Thm)

If A is Hermitian, then there exists a unitary matrix U s.t.

$$A = U D U^H$$

where D is a diagonal matrix.

proof $A \in \mathbb{C}^{n \times n}$, then A has the Schur decomposition as

$$A = U T U^H$$

Since A is Hermitian, $A = A^H$, then

$$T^H = (U^H A U)^H = U^H A^H U = U^H A U = T$$

Because T is upper triangular, so T must be diagonal with $t_{ii} = \bar{t}_{ii}$, for $i=1, \dots, n$.

Hence T must be a real-diagonal matrix.

Example

$$A = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 4 & 0 \\ 1 & 0 & 2 \end{bmatrix} \quad . \quad B = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 1 & 2 \end{bmatrix}$$

Find the Schur's decomposition of A and B .

$$\text{Eigenvalues: } \lambda_1 = 4, \quad \lambda_2 = \lambda_3 = 2$$

Eigenvectors: For $\lambda = 4$

$$(A - 4I) \vec{v} = \vec{0}$$

$$\begin{bmatrix} -2 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$(B - 4I) \vec{x}_1 = \vec{0}$$

$$\text{rank 2 matrix!} \quad \begin{bmatrix} -2 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & -2 \end{bmatrix} \begin{bmatrix} d \\ e \\ f \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$(A - 4I)\vec{v}_1 = \vec{0}$$

rank 2 → nullity 1

$$\begin{bmatrix} -2 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & -2 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{cases} a=0 \\ c=0 \\ b \in \mathbb{R} \end{cases} \rightarrow \vec{v}_1 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \alpha, \alpha \neq 0$$

$$E_{\lambda_1=4} = \text{span}\left\{\begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}\right\}$$

rank 2
nullity 1

$$\begin{bmatrix} -2 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & -2 \end{bmatrix} \begin{bmatrix} d \\ e \\ f \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{cases} d=0 \\ e=2f \\ f \in \mathbb{R} \end{cases} \rightarrow \vec{x}_1 = \begin{pmatrix} 0 \\ 2 \\ 1 \end{pmatrix} \alpha, \alpha \neq 0$$

$$E_{\lambda_2=4} = \text{span}\left\{\begin{pmatrix} 0 \\ 2 \\ 1 \end{pmatrix}\right\}$$

Eigenvectors for $\lambda=2$

$$(A - 2I)\vec{v}_2 = \vec{0}$$

rank 2
nullity 1

$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 1 & 0 & 0 \end{bmatrix} \vec{v}_2 = \vec{0}$$

$$\vec{v}_2 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \beta, \beta \neq 0$$

$$E_{\lambda_3=2} = \text{span}\left\{\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}\right\}$$

rank 1
nullity 2

$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 1 & 0 \end{bmatrix} \vec{x}_2 = \vec{0}$$

$$\rightarrow \vec{x}_2 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \beta + \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \delta, \beta, \delta \text{ are not both 0}$$

$$E_{\lambda_B=2} = \text{span}\left\{\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}\right\}$$

(Notice that A is deficient and B is diagonalizable.)

Schur's decomp : $A = U T U^H$ ① Form U ② Compute $T = U^H A U$

Take B as an example :

① Finding U : Matrix B has $\begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$, $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$, $\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$ linearly independent eigenvectors

Let $\vec{u}_1 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$, since $\vec{u}_1 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$, $\vec{u}_1 \perp \vec{u}_2$ and $\|\vec{u}_1\| = \|\vec{u}_2\| = 1$

$$\text{take } \vec{v}_3 = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} - \underbrace{\text{proj}_{\vec{u}_1} \vec{v}_3}_{\vec{x}_3} - \underbrace{\text{proj}_{\vec{u}_2} \vec{v}_3}_{\vec{u}_3} = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} - \frac{\vec{x}_3^T \vec{u}_1}{\|\vec{u}_1\|^2} \vec{u}_1 - \frac{\vec{x}_3^T \vec{u}_2}{\|\vec{u}_2\|^2} \vec{u}_2$$

$$\vec{v}_3 = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} - \vec{0} - \frac{1}{1} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\text{let } \vec{u}_3 = \frac{1}{\|\vec{v}_3\|} \vec{v}_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$$\text{let } U = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix},$$

$$T = U^H B U = \underbrace{\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}}_{U^H} \underbrace{\begin{bmatrix} 2 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 2 \end{bmatrix}}_{B} \underbrace{\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}}_{U}$$

$$= \begin{bmatrix} 2 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 4 & 0 \end{bmatrix} U$$

$$= \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & 0 & 4 \end{bmatrix} \quad \leftarrow \text{upper triangular}$$

$$= \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & 0 & 4 \end{bmatrix} \quad \leftarrow \text{upper triangular}$$

The Schur decomposition of A : left as exercise

$$\forall A \in \mathbb{C}^{n \times n} \xrightarrow{\text{Schur decomp}} A = U T U^H$$

When
A
Hermitian
 $A = A^H$

\uparrow unitary \uparrow upper triangular

A is unitarily similar to T .

$$\xrightarrow{\text{spectral decomp.}}$$

$$A = U D U^H$$

\uparrow unitary \uparrow diagonal

$$\xrightarrow{A \in \mathbb{R}^{n \times n}}$$

$$A = Q D Q^T$$

\uparrow orthogonal

$A = A^T$
real symmetric

Diagonalization: $\mathbb{R}^{n \times n}$

$$\forall A \in \mathbb{R}^{n \times n}$$

$$\xrightarrow{\text{diagonalizable}}$$

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

\uparrow nonsingular \uparrow diagonal

deficient

$$A = C J C^{-1}$$

\uparrow nonsingular \uparrow Jordan form.

SVD

(Singular Value decomp)

$$A_{m \times n} = U \Sigma_{m \times n} V^T$$

$(m \neq n)$

where U and V are both orthogonal matrices.

We name the diagonal entries of Σ as "singular values" of A .

WLOG, we denote $m > n$

$$\begin{bmatrix} A \\ \vdots \\ A \end{bmatrix}_{m \times n} = \begin{bmatrix} U \\ \vdots \\ U \end{bmatrix}_{m \times n} \begin{bmatrix} \Sigma_1 & & \\ & \ddots & \\ & & \Sigma_n \\ & & 0_{(m-n) \times n} \end{bmatrix}_{m \times n} \begin{bmatrix} V^T \\ \vdots \\ V^T \end{bmatrix}_{n \times n}$$

where the columns of U are eigenvectors of $A A^T$

the columns of V are eigenvectors of $A^T A$

the diagonal entries of Σ are $\sigma_i = \sqrt{\lambda_i}$, with λ_i as eigenvalues of $A^T A$.

$$\begin{aligned} \text{Consider } A^T A &= V \Sigma^T U^T U \Sigma V^T \\ &= V \Sigma^T \Sigma V^T \\ &= V \begin{bmatrix} \sigma_1^2 & & \\ & \ddots & \\ & & \sigma_n^2 \end{bmatrix}_{n \times n} V^T \quad \rightarrow \Sigma^T \Sigma = \Sigma_1^T \Sigma_1 + 0_{n \times (m-n)} 0_{(m-n) \times n} \\ &\quad \text{orthogonal} \quad \text{diagonal} \\ &\quad \text{Since } \Sigma_1 \text{ diagonal} \quad \Sigma_1^T \Sigma_1 = \begin{bmatrix} \sigma_1^2 & & \\ & \sigma_2^2 & \\ & & \ddots & \sigma_n^2 \end{bmatrix}_{n \times n} \end{aligned}$$

$A^T A$ is symmetric and diagonalized by an orthogonal matrix V .

Thus, the columns of V are eigenvectors of $A^T A$

the diagonal entries of Σ_1^2 are eigenvalues of $A^T A$

$$\begin{aligned} \text{Consider } A A^T &= U \Sigma V^T V \Sigma^T U^T \\ &= U \Sigma \Sigma^T U^T \\ &= U \begin{bmatrix} \Sigma_1^2 & 0 \\ 0 & 0 \end{bmatrix}_{n \times n} U^T \quad \rightarrow \Sigma \Sigma^T = \begin{bmatrix} \Sigma_1 & & \\ & \ddots & \\ & & 0_{(m-n) \times n} \end{bmatrix}_{m \times m} \begin{bmatrix} \Sigma_1^T & & \\ & \ddots & \\ & & 0^T \end{bmatrix}_{n \times n} \\ &\quad \text{orthogonal} \quad \text{diagonal} \\ &\quad \text{Since } \Sigma_1 \text{ diagonal} \quad \Sigma_1 \Sigma_1^T = \begin{bmatrix} \Sigma_1 \Sigma_1^T & 0_{n \times (m-n)} \\ 0_{(m-n) \times n} \Sigma_1^T & 0_{(m-n) \times (m-n)} \end{bmatrix}_{m \times m} \\ &\quad \Sigma_1 \Sigma_1^T = \begin{bmatrix} \sigma_1^2 & & \\ & \sigma_2^2 & \\ & & 0 \end{bmatrix}_{n \times n} \end{aligned}$$

$A A^T$ is symmetric and diagonalized by an orthogonal matrix U .

Thus, the columns of U are eigenvectors of $A A^T$

the diagonal entries of Σ_1^2 are eigenvalues of $A A^T$