

Remarks :

① For $A \in \mathbb{R}^{n \times n}$, eigenvectors are not unique for a given eigenvalue.

Example: If $A \in \mathbb{R}^{3 \times 3}$ satisfies $A \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = 2 \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$

$\forall a \neq 0$, $a \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} a \\ a \\ a \end{bmatrix}$ is an eigenvector of A for the eigenvalue $\lambda = 2$.

② $A \in \mathbb{R}^3$, $V = [\vec{v}_1 \ \vec{v}_2 \ \vec{v}_3]$, $D = \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix}$

diagonalization $A = VDV^{-1}$

$\Leftrightarrow A V = V D$ for an invertible V

$\Leftrightarrow A [\vec{v}_1 \ \vec{v}_2 \ \vec{v}_3] = [\vec{v}_1 \ \vec{v}_2 \ \vec{v}_3] \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix}$

$$\Leftrightarrow \begin{cases} A \vec{v}_1 = d_1 \vec{v}_1 \\ A \vec{v}_2 = d_2 \vec{v}_2 \\ A \vec{v}_3 = d_3 \vec{v}_3 \end{cases}$$

$\Leftrightarrow A$ has n linearly independent eigenvectors

$\Leftrightarrow \begin{cases} A \text{ has } n \text{ eigenvalues including repeated ones} \\ \text{For all eigenspace: Geo Mul} = \text{Alg Mul} \end{cases}$

③ If A is diagonalizable, then there are many matrices V, D satisfying $A = VDV^{-1}$

$$A [\vec{v}_1 \vec{v}_2 \vec{v}_3] = [\vec{v}_1 \vec{v}_2 \vec{v}_3] \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix}$$

$$A [\vec{v}_2 \vec{v}_3 \vec{v}_1] = [\vec{v}_2 \vec{v}_3 \vec{v}_1] \begin{bmatrix} d_2 & 0 & 0 \\ 0 & d_3 & 0 \\ 0 & 0 & d_1 \end{bmatrix}$$

$$A [\vec{v}_2 \vec{v}_3 \vec{v}_1] = [\vec{v}_2 \vec{v}_3 \vec{v}_1] \begin{bmatrix} d_2 & 0 & 0 \\ 0 & d_3 & 0 \\ 0 & 0 & d_1 \end{bmatrix}$$

④ Basis vectors for each eigen-space are not unique

⑤ Any non-zero vector in eigen-space is an eigenvector.

The following real matrices are always diagonalizable :

① Real Symmetric Matrix $A = A^T$ $\begin{pmatrix} 1 & 2 & 3 & 4 \\ 2 & 5 & 6 & 7 \\ 3 & 6 & 8 & 9 \\ 4 & 7 & 9 & 10 \end{pmatrix}$

② Real Skew-Symmetric $A^T = -A$ $\begin{pmatrix} 0 & 1 & 2 \\ -1 & 0 & 3 \\ -2 & -3 & 0 \end{pmatrix}$

③ Normal Matrix $AA^T = A^TA$

Theorem If $A \in \mathbb{R}^{n \times n}$ satisfies $A = A^T$, then

- ① A has n real eigenvalues including repeated ones
- ② A has n orthogonal real eigenvectors, i.e.,
all eigenspaces are orthogonal to one another.

Proof: ① $\det(A - \lambda I)$ has n complex roots
including repeated ones.

Each distinct eigenvalue $\lambda = a+ib$ has at least one
eigenvector \vec{v}

$$A\vec{v} = \lambda \vec{v} \quad \overline{a+ib} = a-ib$$

Take conjugate:

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

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

$$(A = A) \quad A\overline{\vec{v}} = \overline{\lambda}\overline{\vec{v}} \quad (A^T)^T = B^T A^T$$

Take transpose:

$$\overline{\vec{v}}^T A^T = \overline{\vec{v}}^T \overline{\lambda}$$

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

$$\overline{\vec{v}}^T A \vec{v} = \overline{\vec{v}}^T \overline{\lambda} \vec{v} \quad (\times)$$

$$A\vec{v} = \lambda\vec{v} \Rightarrow \vec{v}^T A \vec{v} = \vec{v}^T \lambda \vec{v} \quad (***)$$

$$\begin{matrix} (*) \\ (***) \end{matrix} \Rightarrow \vec{v}^T \bar{\lambda} \vec{v} = \bar{\lambda}^T \lambda \vec{v}$$

$$\Rightarrow \bar{\lambda} \vec{v}^T \vec{v} = \lambda \vec{v}^T \vec{v}$$

$$\Rightarrow (\bar{\lambda} - \lambda) \underbrace{\vec{v}^T \vec{v}}_{\downarrow} = 0$$

$$\begin{bmatrix} \bar{v}_1 & \bar{v}_2 & \bar{v}_3 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} = \bar{v}_1 v_1 + \bar{v}_2 v_2 + \bar{v}_3 v_3$$

$$= |v_1|^2 + |v_2|^2 + |v_3|^2 > 0$$

$$\Rightarrow \bar{\lambda} = \lambda \Rightarrow \lambda \in \mathbb{R}$$

② We already know n complex eigenvalues are real, so all eigenvectors are real.

Want to show $A\vec{v}_1 = \lambda_1 \vec{v}_1$

$$\left. \begin{array}{l} A\vec{v}_2 = \lambda_2 \vec{v}_2 \\ \lambda_1 \neq \lambda_2 \end{array} \right\} \Rightarrow \vec{v}_1 \perp \vec{v}_2$$

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

Take dot product with \vec{v}_2

$$\langle A\vec{v}_1, \vec{v}_2 \rangle = \langle \lambda_1 \vec{v}_1, \vec{v}_2 \rangle$$

$$(\vec{v}_2^T A) \vec{v}_1 = \lambda_1 \langle \vec{v}_1, \vec{v}_2 \rangle$$

$$(ABC)^T = C^T B^T A^T$$

$$(A^T \vec{v}_2)^T \vec{v}_1 = \lambda_1 \langle \vec{v}_1, \vec{v}_2 \rangle$$

$$A^T = A$$

$$(A \vec{v}_2)^T \vec{v}_1 = \lambda_1 \langle \vec{v}_1, \vec{v}_2 \rangle$$

$$(\lambda_2 \vec{v}_2)^T \vec{v}_1 = \lambda_1 \langle \vec{v}_1, \vec{v}_2 \rangle$$

$$\lambda_2 \vec{v}_2^T \vec{v}_1 = \lambda_1 \langle \vec{v}_1, \vec{v}_2 \rangle$$

$$\lambda_2 \langle \vec{v}_1, \vec{v}_2 \rangle = \lambda_1 \langle \vec{v}_1, \vec{v}_2 \rangle$$

$$(\lambda_1 - \lambda_2) \langle \vec{v}_1, \vec{v}_2 \rangle = 0$$

$$\Rightarrow \langle \vec{v}_1, \vec{v}_2 \rangle = 0.$$

Example: $A = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}$

Find 3 orthonormal eigenvectors of A.

$$\begin{aligned} \text{Sol: } |A - \lambda I| &= \begin{vmatrix} -\lambda & 1 & 1 \\ 1 & -\lambda & 1 \\ 1 & 1 & -\lambda \end{vmatrix} \\ &= -\lambda(\lambda^2 - 1) - (-\lambda - 1) + (1 + \lambda) \\ &= -\lambda(\lambda + 1)(\lambda - 1) + (\lambda + 1) + (\lambda + 1) \\ &= (\lambda + 1)[-\lambda(\lambda - 1) + 1 + 1] \\ &= (\lambda + 1)[- \lambda^2 + \lambda + 2] \\ &= -(\lambda + 1)^2(\lambda - 2) \end{aligned}$$

$$\lambda_1 = -1, \lambda_2 = 2$$

① Plug in $\lambda_1 = -1$ into $(A - \lambda I) \vec{v} = \vec{0}$

$$\left[\begin{array}{ccc|c} 1 & 1 & 1 & 0 \\ 1 & 1 & 1 & 0 \\ 1 & 1 & 1 & 0 \end{array} \right]$$

$$\Rightarrow \left[\begin{array}{ccc|c} 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

$$\begin{cases} v_2 = s \\ v_3 = t \end{cases} \Rightarrow v_1 = -s - t \Rightarrow \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix} = \begin{pmatrix} -s-t \\ s \\ t \end{pmatrix}$$

$$= s \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} + t \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} \quad \vec{u}_1 \quad \vec{u}_2$$

\Rightarrow Eigen-Space is Span $\left\{ \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} \right\}$

But two basis eigen-vectors are not orthogonal.

Apply Gram-Schmidt Procedure :

$$\vec{v}_1 = \vec{u}_1 = \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}$$

$$\vec{v}_2 = \vec{u}_2 - \frac{\langle \vec{v}_1, \vec{u}_2 \rangle}{\|\vec{v}_1\|^2} \vec{v}_1$$

$$= \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} - \frac{1}{2} \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}$$

$$= \begin{pmatrix} -\frac{1}{2} \\ -\frac{1}{2} \\ 1 \end{pmatrix}$$

Verify $A\vec{v}_2 = H\vec{v}_2$

$$\begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix} \begin{pmatrix} -\frac{1}{2} \\ -\frac{1}{2} \\ 1 \end{pmatrix} = \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \\ -1 \end{pmatrix}$$

② Plug in $\lambda = 2$ into $(A - \lambda I) \vec{v} = \vec{0}$

to find $\vec{v}_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$

So we get orthogonal eigenvectors

$$\vec{v}_1 = \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}, \vec{v}_2 = \begin{pmatrix} -\frac{1}{2} \\ -\frac{1}{2} \\ \frac{\sqrt{2}}{2} \end{pmatrix}, \vec{v}_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

and orthonormal eigenvectors:

$$\vec{w}_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \\ 0 \end{pmatrix}$$

$$\vec{w}_2 = \frac{1}{\sqrt{\frac{3}{2}}} \begin{pmatrix} -\frac{1}{2} \\ -\frac{1}{2} \\ 1 \end{pmatrix} = \begin{pmatrix} -\frac{1}{2}\sqrt{\frac{2}{3}} \\ -\frac{1}{2}\sqrt{\frac{2}{3}} \\ \sqrt{\frac{2}{3}} \end{pmatrix}$$

$$\vec{w}_3 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} \frac{\sqrt{3}}{3} \\ \frac{\sqrt{3}}{3} \\ \frac{\sqrt{3}}{3} \end{pmatrix}$$

$$\text{Use } V = [\vec{w}_1 \ \vec{w}_2 \ \vec{w}_3] \quad D = \begin{bmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$

$$\text{Then } A = V D V^{-1}$$

$$\text{and } V^{-1} = V^T$$

$$V^T V = \boxed{\begin{array}{|c|} \hline \end{array}} \boxed{\begin{array}{|c|c|c|} \hline \end{array}} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Theorem If $A \in \mathbb{R}^{n \times n}$ is skew-symmetric ($A^T = -A$),
then

- ① A has n purely imaginary eigenvalues including repeated ones

② A has n orthogonal complex eigenvectors, i.e.,
all eigenspaces are orthogonal to one another.

Example: $A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$ $\lambda_1 = i$ $\vec{v}_1 = \begin{pmatrix} -i \\ 1 \end{pmatrix} = \overline{\vec{v}_2}^T \vec{v}_1$
 $\lambda_2 = -i$ $\vec{v}_2 = \begin{pmatrix} i \\ 1 \end{pmatrix} = [-i \ 1] \begin{pmatrix} -i \\ 1 \end{pmatrix}$

For two complex vectors $\vec{u}, \vec{v} \in \mathbb{C}^n$, the dot product is defined as $= (-i)^2 + 1 = 0$

$$\langle \vec{u}, \vec{v} \rangle = \overline{\vec{v}}^T \vec{u} = [\bar{v}_1 \ \bar{v}_2 \ \bar{v}_3] \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} = \sum_{i=1}^n u_i \bar{v}_i$$

$$\text{So } \langle \vec{u}, \vec{v} \rangle = \overline{\langle \vec{v}, \vec{u} \rangle}$$

We define $\vec{u} \perp \vec{v}$ as $\langle \vec{u}, \vec{v} \rangle = 0$.

$$\langle \vec{v}_1, \vec{v}_2 \rangle = \overline{\vec{v}_2}^T \vec{v}_1 = [-i \ 1] \begin{pmatrix} -i \\ 1 \end{pmatrix} = (-i)^2 + 1 = 0.$$

Theorem If $A \in \mathbb{R}^{n \times n}$ is normal ($A A^T = A^T A$)

then

- ① A has n complex eigenvalues including repeated ones
- ② A has n orthogonal complex eigenvectors, i.e., all eigenspaces are orthogonal to one another.

Example: $A = \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix}$ is normal

$$\begin{aligned} A^T A &= \begin{pmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{pmatrix} \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \end{aligned}$$

$$\begin{aligned} A A^T &= \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix} \begin{pmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \end{aligned}$$

Ex: True or False

- ① If A is real symmetric, then A is normal.
- ② If A is skew-symmetric, then A is normal.

$$\begin{aligned} A^T &= -A \Rightarrow \left. \begin{aligned} A A^T &= A \cdot (-A) = -A^2 \\ A^T A &= (-A) \cdot A = -A^2 \end{aligned} \right\} \end{aligned}$$