

## \* Choosing a good basis for the matrix of linear operator

Recall:  $[T(v)]_B = [T]_{B(B)} [v]_B$

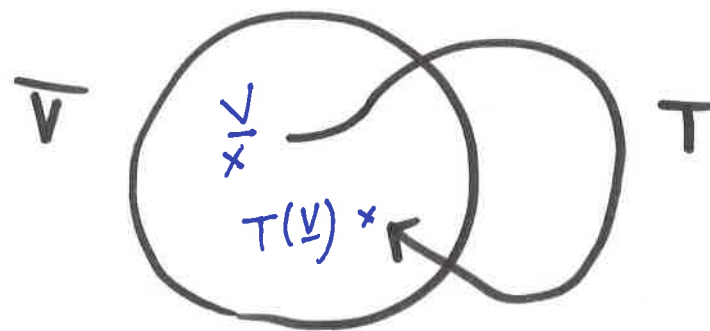
$\Phi_{B,D}^{-1} [T(v)]_D = [T]_{D(D)} [v]_D$

$[T]_{D(D)} = \underbrace{\Phi_{D,B}}_{//} [T]_{B(B)} \underbrace{\Phi_{B,D}}_{||}$

$\left[ \begin{bmatrix} b_1 \end{bmatrix}_D \begin{bmatrix} b_2 \end{bmatrix}_D \cdots \begin{bmatrix} b_n \end{bmatrix}_D \right]$

$\left[ \begin{bmatrix} d_1 \end{bmatrix}_B \begin{bmatrix} d_2 \end{bmatrix}_B \cdots \begin{bmatrix} d_n \end{bmatrix}_B \right]$

$\Phi_{D,B} = \Phi_{B,D}^{-1}, \quad \Phi_{B,D} = \Phi_{D,B}^{-1}, \quad \Phi_{B,D} \Phi_{D,B} = \Phi_{D,B} \Phi_{B,D} = I$



basis:  $B = \{b_1, b_2, \dots, b_n\}$

basis:  $D = \{d_1, d_2, \dots, d_n\}$

diagonalization of matrix  
(Recall:  $D = P^{-1} A P$ )

- At first, we have some basis  $B$ , corresponding P.129 to which  $[T]_{B(B)}$  is probably not in a simple form.
- It would be nice if we can find some basis  $D$   $\Rightarrow [T]_D$  is in some kind of simple form.

(such that)

$\hookrightarrow$  diagonal

- We have learned that the trick is to find the eigen-values/vectors of  $[T]_B$ .
- We need enough l.i. eigenvectors to obtain a diagonal  $[T]_D$ .  
from  $[T]_B$

$\downarrow$   
consisting of the e. values of  $[T]_B$

• Recheck:  $[T]_D = \underline{\underline{\Phi}}_{DB} [T]_B \underline{\underline{\Phi}}_{BD}$

$\xrightarrow{\text{red arrow}} = \underline{\underline{\Phi}}_{BD}^{-1}$

In addition to making  $[T]_B$  look simple, we may even further seek to make  $\underline{\underline{\Phi}}_{DB} / \underline{\underline{\Phi}}_{BD}$  look simple or easy to handle.

• Q: What <sup>^</sup>is the so-called simple/easy/useful/helpful (more) exactly (speaking) property of  $\underline{\underline{\Phi}}_{DB} / \underline{\underline{\Phi}}_{BD}$ ?

Ans: It would be very nice if the column vectors of  $\underline{\underline{\Phi}}_{BD}$  are o.n. vectors.

$\Rightarrow \underline{\underline{\Phi}}_{BD}^T \underline{\underline{\Phi}}_{BD} = \underline{\underline{I}}$   
 $\rightarrow$  orthonormal

• We also know that  $\underline{\underline{\Phi}}_{DB} \underline{\underline{\Phi}}_{BD} = \underline{\underline{I}}$

$\left\{ \begin{array}{l} \underline{\underline{\Phi}}_{DB} = \underline{\underline{\Phi}}_{BD}^{-1} \end{array} \right.$

$\rightarrow \underline{\underline{\Phi}}_{BD}^{-1} = \underline{\underline{\Phi}}_{BD}^T$



- Recall: the def. of "orthogonal matrix":

P.131

(in the lecturer's opinion, orthonormal matrix would have been a better terminology)

$$\underline{\underline{Q^T Q}} = \underline{\underline{I}} \quad (\leadsto \underline{\underline{Q Q^T}} = \underline{\underline{I}})$$

physical interpretation: column vectors of  $\underline{\underline{Q}}$  are o.n.   
 column vectors of  $\underline{\underline{Q^T}}$  ( $\equiv$  row vectors of  $\underline{\underline{Q}}$ ) are o.n.

- Summary: The columns/rows of an orthogonal matrix are o.n. vectors.


- Q: Can we find the afore-mentioned "simple"  $\underline{\underline{\Phi}}_{BD}$  for a given  $[T]_B$ ?

Ans: To be answered later.

• If orthogonal matrix  $\underline{\underline{\Phi}}_{BD}$  does exist,

P.132

then  $[T]_D = \underline{\underline{\Phi}}_{BD}^T [T]_B \underline{\underline{\Phi}}_{BD}$



orthogonal diagonalization of matrix

(Recall:  $\underline{\underline{D}} = \underline{\underline{Q}}^T \underline{\underline{A}} \underline{\underline{Q}}$ )

• Q: What is the advantage of a diagonal  $[T]_D$ ?

Ans: The linear operation (performed by a Lop) can be performed component-wise (i.e. component by component), by simply scaling-up each component vector, which is an eigenvector, by the corresponding eigen-value.

• Ex.  $T\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) \triangleq \begin{bmatrix} x_1 + x_2 \\ -2x_1 + 4x_2 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 1 \\ -2 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

P.133

$[T]_B$ , where  $B = \left\{ \underbrace{\begin{bmatrix} 1 \\ 0 \end{bmatrix}}_{b_1}, \underbrace{\begin{bmatrix} 0 \\ 1 \end{bmatrix}}_{b_2} \right\}$

$\Phi_{DB} = P^{-1} = \begin{bmatrix} 2 & -1 \\ -1 & 1 \end{bmatrix}$

• e. values

e. vectors

$\lambda_1 = 2 \rightarrow \underline{v}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

$\lambda_2 = 3 \rightarrow \underline{v}_2 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

$\Phi_{BD} = P$

$\underline{P} \triangleq [\underline{v}_1 \ \underline{v}_2] = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}$

$\rightarrow = [\underline{d}_2]_B$

•  $D \triangleq \left\{ 1 \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} + 1 \cdot \begin{bmatrix} 0 \\ 1 \end{bmatrix}, 1 \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} + 2 \cdot \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\} = \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \end{bmatrix} \right\} = [\underline{d}]_B$

• (Ex) Let us consider  $\underline{v} = \begin{bmatrix} 3 \\ -1 \end{bmatrix}$ , and

then check its image after the application of  $T$ .

$T\left(\begin{bmatrix} 3 \\ -1 \end{bmatrix}\right) = \begin{bmatrix} 1 & 1 \\ -2 & 4 \end{bmatrix} \begin{bmatrix} 3 \\ -1 \end{bmatrix} = \begin{bmatrix} 2 \\ -10 \end{bmatrix} \quad \text{--- } (@1)$

• If we want to make  $[T]_{@}$  simple, then we adopt  $[T]_D$   
(at the cost of more effort in finding  $[\underline{v}]_D$ )

• Let us find  $\begin{pmatrix} 3 \\ -1 \end{pmatrix}_D : \rightsquigarrow =$

$$\begin{bmatrix} 3 \\ -1 \end{bmatrix} = \alpha \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \beta \cdot \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}^{-1}}_{\text{red wavy line}} \begin{bmatrix} 3 \\ -1 \end{bmatrix} = \underbrace{\begin{bmatrix} 2 & -1 \\ -1 & 1 \end{bmatrix}}_{\text{red wavy line}} \begin{bmatrix} 3 \\ -1 \end{bmatrix} = \begin{bmatrix} 7 \\ -4 \end{bmatrix}$$

$$\begin{bmatrix} V \\ - \end{bmatrix}_B$$

This is  $\underline{\Phi}_{DB}$ , which can be calculated in advance

$$\cdot \underline{T(V)}_D = \underline{T}_D \cdot \underline{V}_D = \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 7 \\ -4 \end{bmatrix} = \begin{bmatrix} 14 \\ -12 \end{bmatrix}$$

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

$$\underline{P}^{-1} \underline{T}_B \underline{P} = \begin{bmatrix} 2 & -1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -2 & 4 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} \textcircled{2} & 0 \\ 0 & \textcircled{3} \end{bmatrix}$$

$\lambda_2$

$\lambda_2$

$$\cdot T(V) = 14 \cdot \underline{d}_1 + (-12) \cdot \underline{d}_2 = 14 \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix} - 12 \cdot \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 2 \\ -10 \end{bmatrix} \text{ --- (@2)}$$

• check:  $(@1) = (@2)$



# \* Orthogonal diagonalizability of matrices P.135

• Thm : Consider  $\underline{\underline{A}} \in \mathbb{R}^{n \times n}$  (i.e. real square matrix).

↑  
**N.B.**

$\underline{\underline{A}}$  is orthogonally diagonalizable iff  $\underline{\underline{A}}$  is symmetric.

↓ $\Delta$

$$\underline{\underline{D}} = \underline{\underline{Q}}^T \underline{\underline{A}} \underline{\underline{Q}} \rightarrow \text{diagonal}$$

↓  
orthogonal matrix

↓ $\Delta$

$$\underline{\underline{A}}^T = \underline{\underline{A}}$$

• Prf : **Beyond the scope of this course.**

• The consideration/discussions on  $\mathbb{R}^{n \times n}$  matrices can be generalized to  $\mathbb{C}^{n \times n}$  matrices.

↑  
complex numbers  $a+j \cdot b$   
(i)  $= \sqrt{-1}$



# $\underline{D} = \underline{P}^{-1} \underline{A} \underline{P}$ : diagonalization

P. 136

orthogonal diag.

(Scenario:  $\underline{A}$  : real)

$$\underline{D} = \underline{P}^{-1} \underline{A} \underline{P}, \text{ where } \underline{P}^{-1} = \underline{P}^T$$

\*  $\underline{A}$  is diagonalizable  
orthogonally

iff  $\underline{A}$  is symmetric.

transpose  $\rightarrow$

$$\underline{A}^T = \underline{A}$$

$\underline{P}$  is orthogonal

unitary diag.

(Scenario:  $\underline{A}$  : complex)

$$\underline{D} = \underline{P}^{-1} \underline{A} \underline{P}, \text{ where } \underline{P}^{-1} = \underline{P}^H$$

\*  $\underline{A}$  is unitarily diagonalizable

iff  $\underline{A}$  is Hermitian.

transpose  $\rightarrow$  conj. symmetric

combined  $\rightarrow$

$$\underline{A}^H = \underline{A}$$

Hermitian (op.)

$\underline{P}$  is unitary

• A: square matrix

P. 137

real and symmetric  $\longleftrightarrow$  orthogonally diagonalizable

(complex and) Hermitian  $\overset{\checkmark}{\longleftrightarrow}$  unitarily diagonalizable

$\swarrow \checkmark \quad \searrow \times$   
(complex and) normal

• def normal matrix:  $\underline{\underline{A}}^H \underline{\underline{A}} = \underline{\underline{A}} \underline{\underline{A}}^H$

• cf: unitary matrix:  $\underline{\underline{A}}^H \underline{\underline{A}} = \underline{\underline{A}} \underline{\underline{A}}^H = \underline{\underline{I}}$

• Thm The e.values of a Hermitian matrix are real numbers.

P. 138

Prf •  $A \underline{x} = \lambda \underline{x}$  — (1)

•  $\underline{x}^H \cdot \text{Eq. (1)} \Rightarrow \underline{x}^H A \underline{x} = \underline{x}^H \cdot \lambda \underline{x} = \lambda \underline{x}^H \underline{x} = \lambda \|\underline{x}\|^2$  — (2)

•  $(\text{Eq. (2)})^H = (\underline{x}^H A \underline{x})^H = (\lambda \cdot \|\underline{x}\|^2)^H = \lambda^* \cdot \|\underline{x}\|^2$  — (3)

$\hookrightarrow \text{real}$

$(\underline{PQ})^H = \underline{Q}^H \underline{P}^H \rightarrow \underline{x}^H A^H \underline{x}$

$\underline{x}^H A \underline{x} = \lambda^* \cdot \|\underline{x}\|^2$

$A$  is Hermitian  $\rightarrow A = A^H$

•  $\left. \begin{matrix} (2) \\ (3) \end{matrix} \right\} \Rightarrow \lambda = \lambda^* \Rightarrow \lambda \text{ is real} \neq$

Prf real and symmetric  
 $\downarrow$  implies  
Hermitian

• Corollary The e.values of a symmetric matrix with real entries are real numbers.

Thm Let  $\underline{A}$  be a Hermitian matrix.

P.139

The e.vectors of  $\underline{A}$  wrt different e. values are orthogonal.

(IP:  $\langle \underline{a}, \underline{b} \rangle = \left\langle \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}, \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} \right\rangle \triangleq \bar{a}_1 b_1 + \bar{a}_2 b_2 + \dots + \bar{a}_n b_n$   
 $\triangleq [\bar{a}_1 \bar{a}_2 \dots \bar{a}_n] \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} = \underline{a}^H \underline{b}$ )

Prf If  $\underline{x}_1$  and  $\underline{x}_2$  are eigenvectors wrt distinct eigenvalues  $\lambda_1$  and  $\lambda_2$ , respectively, then

$$\cdot (\underline{A} \underline{x}_1)^H \underline{x}_2 = \underline{x}_1^H \underline{A}^H \underline{x}_2 = \underline{x}_1^H \underline{A} \underline{x}_2 = \lambda_2 \underline{x}_1^H \underline{x}_2 \quad \text{--- (#1)}$$

"  $\leftarrow$  Hermitian "  $\lambda_1 \underline{x}_1$

$$\cdot (\underline{A} \underline{x}_1)^H \underline{x}_2 = (\underline{x}_2^H (\underline{A} \underline{x}_1))^H = (\underline{x}_2^H \underline{A} \underline{x}_1)^H = (\lambda_1 \underline{x}_2^H \underline{x}_1)^H = \lambda_1 \underline{x}_1^H \underline{x}_2 \quad \text{--- (#2)}$$

$$\cdot \text{(#1)} - \text{(#2)}: (\lambda_2 - \lambda_1) \underline{x}_1^H \underline{x}_2 = 0 \Rightarrow \underline{x}_1^H \underline{x}_2 = 0$$

$\lambda_2 \neq \lambda_1$

$\underline{x}_1 \perp \underline{x}_2$   $\Delta$



• Ex

$$T\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = \begin{bmatrix} x_1 + x_2 \\ x_1 + 4x_2 \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & 1 \\ 1 & 4 \end{bmatrix}}_{\text{real and symmetric}} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\underline{\underline{A}} \equiv \underline{\underline{[T]_B}}, \text{ where } B \triangleq \left\{ \underbrace{\begin{bmatrix} 1 \\ 0 \end{bmatrix}}_{\underline{\underline{b_1}}}, \underbrace{\begin{bmatrix} 0 \\ 1 \end{bmatrix}}_{\underline{\underline{b_2}}} \right\}$$

real and symmetric

P.140

• eigen-analysis:

$$\lambda_1 = \frac{5 - \sqrt{13}}{2} (\approx .6972)$$

e. vector

$$\begin{bmatrix} 1 \\ \frac{3 - \sqrt{13}}{2} \end{bmatrix}$$

$$\approx \begin{bmatrix} 1 \\ -0.3028 \end{bmatrix}$$

↑ normalize

(→ norm  $\approx 1.0448$ )

$$\lambda_2 = \frac{5 + \sqrt{13}}{2} (\approx 4.3028)$$

e. vector

$$\begin{bmatrix} 1 \\ \frac{3 + \sqrt{13}}{2} \end{bmatrix}$$

$$\approx \begin{bmatrix} 1 \\ 3.3028 \end{bmatrix}$$

(→ norm  $\approx 3.4508$ )

"I"

• We choose

$$\underline{\underline{Q}} = [\underline{\underline{d_1}} \quad \underline{\underline{d_2}}] = \begin{bmatrix} .9571 & .2898 \\ -.2898 & .9571 \end{bmatrix}$$

↓ normalize

$$\begin{bmatrix} .2898 \\ .9571 \end{bmatrix} \triangleq \underline{\underline{d_2}}$$

(check:  $\underline{\underline{Q}}^T \underline{\underline{Q}} = \underline{\underline{Q}} \underline{\underline{Q}}^T = \underline{\underline{I}} \leadsto \underline{\underline{Q}}^{-1} = \underline{\underline{Q}}^T$ )

• (Ex) Let us consider  $\underline{v} = \begin{bmatrix} 3 \\ -1 \end{bmatrix}$ , and try to find  $T(\underline{v})$ .

P.141

• By adopting B as the o.b,  $\left( \begin{bmatrix} 3 \\ -1 \end{bmatrix} \cong 3 \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} + (-1) \cdot \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right)$

$$(T(\begin{bmatrix} 3 \\ -1 \end{bmatrix}))_B = [T]_B \cdot \begin{bmatrix} 3 \\ -1 \end{bmatrix}_B = \begin{bmatrix} 1 & 1 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} 3 \\ -1 \end{bmatrix} = \begin{bmatrix} 2 \\ -1 \end{bmatrix}$$

$$\therefore T(\begin{bmatrix} 3 \\ -1 \end{bmatrix}) = 2 \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} + (-1) \cdot \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ -1 \end{bmatrix} \quad \text{--- } (\star 1)$$

• If we adopt D as the o.b.  $(D = \{\underline{d}_1, \underline{d}_2\})$

$$\begin{bmatrix} 3 \\ -1 \end{bmatrix} \xrightarrow{\text{coord. vector wrt. D}} \underbrace{\Phi_{DB}}_{\downarrow} \cdot \begin{bmatrix} \underline{v} \end{bmatrix}_B = \begin{bmatrix} 3 \\ -1 \end{bmatrix}$$

$$= \left\{ \begin{bmatrix} .9571 \\ -.2898 \end{bmatrix}, \begin{bmatrix} .2898 \\ .9571 \end{bmatrix} \right\}$$

$$= \Phi_{BD}^{-1} = \begin{bmatrix} \underline{d}_1 & \underline{d}_2 \end{bmatrix}^{-1} = \underline{Q}^{-1} = \underline{Q}^T = \begin{bmatrix} .9571 & -.2898 \\ .2898 & .9571 \end{bmatrix}$$

$$\underline{Q} = \begin{bmatrix} .9571 & .2898 \\ -.2898 & .9571 \end{bmatrix}$$

$$\cdot \left( \begin{bmatrix} 3 \\ -1 \end{bmatrix} \right)_D = \begin{bmatrix} .9571 & -.2898 \\ .2898 & .9571 \end{bmatrix} \begin{bmatrix} 3 \\ -1 \end{bmatrix} = \begin{bmatrix} 3.1611 \\ -.0877 \end{bmatrix}$$

$$\cdot [T([3])_D] = [T]_D \cdot ([3])_D$$

$$= \begin{bmatrix} \frac{5-\sqrt{13}}{2} & 0 \\ 0 & \frac{5+\sqrt{13}}{2} \end{bmatrix} \begin{bmatrix} 3.1611 \\ -0.0877 \end{bmatrix} = \begin{bmatrix} 2.2040 \\ -0.3775 \end{bmatrix}$$

$$\cdot T([3]) = \overbrace{2.2040}^{\doteq x_1} \cdot \underline{d}_1 + \overbrace{(-0.3775)}^{\doteq x_2} \cdot \underline{d}_2 \quad \left( = \begin{bmatrix} \underline{d}_1 & \underline{d}_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \right)$$

$$= 2.2040 \cdot \begin{bmatrix} .9571 \\ -.2898 \end{bmatrix} - .3775 \cdot \begin{bmatrix} .2898 \\ .9571 \end{bmatrix}$$

$$= \begin{bmatrix} 2 \\ -1 \end{bmatrix} \quad \text{---} \quad (\star 2)$$

$$\cdot \text{Check: } (\star 1) = (\star 2)$$

P. 143

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clc;
clear all;
A = [2+i 0 0; 0 0 3-2*i; 0 2+3*i 0];
[P,D] = eig(A)
P_inv = inv(P)
Pinv_A_P = P_inv*A*P
P_herm = P'
Pherm_A_P = P_herm*A*P

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$$\underline{A} = \begin{bmatrix} 2+\bar{i} & 0 & 0 \\ 0 & 0 & 3-2\bar{i} \\ 0 & 2+3\bar{i} & 0 \end{bmatrix}$$

→ normal  
( $\underline{A}^H \underline{A} = \underline{A} \underline{A}^H$ )

$$\underline{A} = \begin{bmatrix} 2.0000 + 1.0000i & 0.0000 + 0.0000i & 0.0000 + 0.0000i \\ 0.0000 + 0.0000i & 0.0000 + 0.0000i & 3.0000 - 2.0000i \\ 0.0000 + 0.0000i & 2.0000 + 3.0000i & 0.0000 + 0.0000i \end{bmatrix}$$

$$\underline{P} = \begin{bmatrix} 0.0000 + 0.0000i & 0.0000 + 0.0000i & 1.0000 + 0.0000i \\ 0.7071 + 0.0000i & 0.7071 + 0.0000i & 0.0000 + 0.0000i \\ 0.5000 + 0.5000i & -0.5000 - 0.5000i & 0.0000 + 0.0000i \end{bmatrix}$$

$$\underline{D} = \begin{bmatrix} 3.5355 + 0.7071i & 0.0000 + 0.0000i & 0.0000 + 0.0000i \\ 0.0000 + 0.0000i & -3.5355 - 0.7071i & 0.0000 + 0.0000i \\ 0.0000 + 0.0000i & 0.0000 + 0.0000i & 2.0000 + 1.0000i \end{bmatrix}$$

$$\underline{P}_{\text{inv}} = \begin{bmatrix} 0.0000 + 0.0000i & 0.7071 - 0.0000i & 0.5000 - 0.5000i \\ 0.0000 + 0.0000i & 0.7071 + 0.0000i & -0.5000 + 0.5000i \\ 1.0000 + 0.0000i & 0.0000 + 0.0000i & 0.0000 + 0.0000i \end{bmatrix}$$

$$\underline{P}_{\text{inv}} \underline{A} \underline{P} = \begin{bmatrix} 3.5355 + 0.7071i & 0.0000 + 0.0000i & 0.0000 + 0.0000i \\ -0.0000 + 0.0000i & -3.5355 - 0.7071i & 0.0000 + 0.0000i \\ 0.0000 + 0.0000i & 0.0000 + 0.0000i & 2.0000 + 1.0000i \end{bmatrix}$$

$$\underline{P}_{\text{herm}} = \begin{bmatrix} 0.0000 + 0.0000i & 0.7071 + 0.0000i & 0.5000 - 0.5000i \\ 0.0000 + 0.0000i & 0.7071 + 0.0000i & -0.5000 + 0.5000i \\ 1.0000 + 0.0000i & 0.0000 + 0.0000i & 0.0000 + 0.0000i \end{bmatrix}$$

$$\underline{P}_{\text{herm}} \underline{A} \underline{P} = \begin{bmatrix} 3.5355 + 0.7071i & 0.0000 - 0.0000i & 0.0000 + 0.0000i \\ 0.0000 + 0.0000i & -3.5355 - 0.7071i & 0.0000 + 0.0000i \\ 0.0000 + 0.0000i & 0.0000 + 0.0000i & 2.0000 + 1.0000i \end{bmatrix}$$

$$\underline{A} \underline{P}_1 = \lambda_1 \underline{P}_1$$

$$\underline{A} \underline{P}_2 = \lambda_2 \underline{P}_2$$

$$\underline{A} \underline{P}_3 = \lambda_3 \underline{P}_3$$

$$\underline{P}^H = \underline{P}^{-1}$$



$\underline{P}$  is unitary.