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Yesterday ...

Every symmetric matrix is  
diagonalizable by an  
orthogonal matrix of eigenvectors!

This is also called the  
spectral decomposition of  
a symmetric matrix.

$$A = Q \Lambda Q^T \xrightarrow{Q^T = Q^{-1}}$$

$\underbrace{\quad}_{\Lambda \text{ diagonal matrix}} \text{ of } \lambda_i.$

Thm A symmetric matrix is  $\Rightarrow$   
 positive definite iff  $\Leftarrow$   
 all the eigenvalues are positive.

Pf: let  $K$  be positive definite.  $\Rightarrow$

By def.  $q(x) = x^T K x > 0$   
 $\forall x \neq 0.$

But  $K$  is symmetric, so

$K = Q \Delta Q^T$ , by spectral  
 decomposition  $\Delta = \begin{pmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \ddots \\ & & & \lambda_n \end{pmatrix}$

$$\begin{aligned} q(x) &= x^T K x = x^T Q \Delta Q^T x \\ &= (Q^T x)^T \Delta (Q^T x) \end{aligned}$$

$$q_b(x) = (Q^T x)^T \Delta (Q^T x)$$

If we let  $y = Q^T x$        $Q^T$  is invertible

$$\Rightarrow x = Qy$$

$$q_b(y) = y^T \Delta y$$

$$q_b(x) = q_b(y) = (y_1, \dots, y_n) \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

$$= \lambda_1 y_1^2 + \lambda_2 y_2^2 + \dots + \lambda_n y_n^2$$

$$= \sum_{i=1}^n \lambda_i y_i^2 > 0$$

If  $y = \tilde{e}_i$  then  $q_b(e_i) = \lambda_i > 0$

$\Leftarrow$  If  $\lambda_i > 0$

$$q_b(x) = y^T \Delta y = \sum \lambda_i y_i^2$$

$$\text{Since } \lambda_i > 0, \quad q_b(x) = \sum \lambda_i y_i^2 > 0$$

□

$$\underline{\text{Ex}} \quad A = \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix}$$

$\lambda_1 = 2, \quad \lambda_2 = 4$

$v_1 = \begin{pmatrix} -1 \\ 1 \end{pmatrix}, \quad v_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$

$4, 2 > 0$   
so  
 $A$  is positive definite!

$\downarrow \text{G-S}$        $\downarrow \text{G-S}$

$\frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 1 \end{pmatrix}, \quad \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$

To make an orthonormal basis  $\{\}$  of eigenvectors for  $A$ , do G-S on each  $v_\lambda$  individually.

Spectral decomposition

$$A = Q \Lambda Q^T$$

$$Q = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \end{pmatrix} \quad \Lambda = \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix}$$

$$\begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \end{pmatrix} \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} \end{pmatrix}$$

$$\begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix} \longrightarrow q(x) = x^T \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix} x$$

$$= (x_1, x_2) \begin{pmatrix} 3 & 1 \\ 1 & 3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$= 3x_1^2 + 2x_1x_2 + 3x_2^2$$

$$\text{ (blue arrow)} = 2x_1^2 + 2x_2^2 + x_1^2 + 2x_1x_2 + x_2^2$$

$$= 2x_1^2 + 2x_2^2 + (x_1 + x_2)^2 > 0$$

On the other hand

$$q(x) = \underbrace{(x_1, x_2)}_{\frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}} \underbrace{\frac{1}{\sqrt{2}} \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix}}_{\frac{1}{2} \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix}} \underbrace{\begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix}}_{\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}}$$

$$= \frac{1}{2} (x_1 + x_2, -x_1 + x_2) \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} x_1 + x_2 \\ -x_1 + x_2 \end{pmatrix}$$

$$\text{Let } y_1 = x_1 + x_2 \quad y_2 = -x_1 + x_2$$

$$q(y) = \frac{1}{2} (y_1, y_2) \begin{pmatrix} 4 & 0 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$$

$$= \frac{1}{2} (4y_1^2 + 2y_2^2) > 0$$

$$q(x) = \frac{1}{2} (4y_1^2 + 2y_2^2) \quad y_1 = x_1 + x_2 \\ y_2 = -x_1 + x_2$$

$$= \frac{1}{2} (4(x_1 + x_2)^2 + 2(-x_1 + x_2)^2) > 0$$

$$\therefore 3x_1^2 + 2x_1x_2 + 3x_2^2 > 0$$

Recall that

$K$  being pos. def

$$\Leftrightarrow \langle x, y \rangle = x^T K y$$

defined an  
inner product

$$f(x, y) \quad \min / \max \quad Df = 0$$

$$Hf = \begin{bmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial y \partial x} & \frac{\partial^2 f}{\partial y^2} \end{bmatrix} \quad \begin{array}{l} \text{if } Hf \\ \text{pos def} \end{array} \Rightarrow \begin{array}{l} \text{min} \\ \text{value} \end{array}$$

Ex Consider ?

$$g(x,y,z) = x^2 + 2xz + y^2 - 2yz \geq 0$$

with  $g(x,y,z) = (x\ y\ z)^T K \begin{pmatrix} x \\ y \\ z \end{pmatrix}$

and find the eigenvalues of  $K$ !

$$K = \begin{pmatrix} x & y & z \\ 1 & 0 & 1 \\ 0 & 1 & -1 \\ 1 & -1 & 0 \end{pmatrix} \quad g(\vec{x}) = \vec{x}^T \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 1 & -1 & 0 \end{pmatrix} \vec{x}$$

Compute eigenvalues  $\rightarrow$

$$\boxed{r\lambda = -1} \quad \text{not positive!}$$

$$\lambda = 1$$

$$\lambda = 2$$

so  $g(x) \neq 0$  for all  $x$ .

$$\lambda = -1 \Rightarrow v = \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix} \quad g(-1, 1, 2) < 0$$

eigenvector

Thm let  $K$  be a symmetric matrix

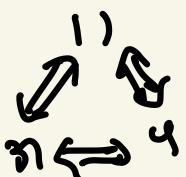
let  $K_i =$  the matrix formed by  
the first  $i$  rows and columns  
of  $K$ .

$$\left( \tilde{e}_x \quad \begin{array}{|c|c|} \hline 1 & 0 \\ 0 & 1 \\ \hline \end{array} \quad \begin{array}{|c|} \hline 1 \\ -1 \\ \hline \end{array} \quad K_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right)$$

Then TFAE

- 1)  $K$  is positive definite
- 2) all eigenvalues of  $K > 0$
- 3) all pivots of  $K$  are positive
- 4)  $\det K_i > 0 \quad \forall i.$  ✓

pf 1)  $\Leftrightarrow$  2) ✓



equiv by  
 $A = LDU$

What if you can't diagonalize?

$A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$  has no diagonalization

what can we say about it?

- Schur decomposition (not common)
- Jordan Canonical Form  
(Jordan decomposition) (more common)
  - more intuitively close to diagonalization
  - "more sophisticated"

## Schur decomposition

If  $A$  is not symmetric, complex eigenvalues / eigenvectors are a possibility.

$A: \mathbb{C}^n \rightarrow \mathbb{C}^n$ , since

$v \in \mathbb{C}^n$  not  $v \in \mathbb{R}^n$

or  $\lambda \in \mathbb{C}$ .

$$z \cdot w = z^T \bar{w}$$

orthogonal matrices don't quite make sense in  $\mathbb{C}^n$ . We need a new concept, called a unitary matrix.

Def: A matrix  $U \in M_{n \times n}(\mathbb{C})$   
is called unitary if

$$U^{-1} = \overline{U^T}.$$

Def:  $\overline{U^T}$  is often called the  
Hermitian of  $U$ .

$$U^H = \overline{U^T} \text{ or } U^+ = \overline{U^T} = \bar{U}^T$$

I've seen  
this

book, I'll use  
this one, NOT  
ADJOINT

Prop  $U$  is unitary  
iff the columns of  $U$   
form an orthonormal basis  
in  $\mathbb{C}^n$ .

Pf: If  $u$  is unitary, then

$$u^{-1} = \overline{u^\top}$$

$$u^{-1}u = I$$

But  $u = (u_1, u_2, \dots, u_n)$

$$u_i \in \mathbb{C}^n$$

$$u^{-1}u = \left( \begin{array}{c} \overline{u_1} \\ \overline{u_2} \\ \vdots \\ \overline{u_n} \end{array} \right) (u_1 \dots u_n) = I$$

need

$$\Leftrightarrow \left\{ \begin{array}{l} \overline{u_i} \cdot u_j = 0 \text{ if } i \neq j \\ \overline{u_i} \cdot u_i = 1 \text{ if } i = j \end{array} \right.$$

$\Leftrightarrow \{u_i\}_{i=1}^n$  form an orthonormal basis  
of  $\mathbb{C}^n$ !

D

$$\underline{\text{Ex}} \quad u = \begin{bmatrix} i & 0 \\ 0 & i \end{bmatrix} = (\text{i.e. } \text{icv})$$

$$\overline{u^T} = \overline{\begin{bmatrix} i & 0 \\ 0 & i \end{bmatrix}} = \begin{bmatrix} -i & 0 \\ 0 & -i \end{bmatrix}$$

$$\overline{u^T} u = \begin{bmatrix} -i & 0 \\ 0 & -i \end{bmatrix} \begin{bmatrix} i & 0 \\ 0 & i \end{bmatrix}$$

$$= \begin{bmatrix} -i \cdot i & 0 \\ 0 & -i \cdot i \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$u^T u = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \neq I$$

i.e.,  $i e_1$  should be orthonormal

$$\|ie_1\| = |i| \|e_1\| = 1 \cdot 1 = 1$$

If we want to generate the spectral decomposition

$$A = \boxed{Q} \Lambda Q^T$$

orth.

to a general matrix.

We need to possibly consider an orthogonal matrix but on  $\mathbb{C}^n$

Orth. +  $\mathbb{C}^n \rightarrow$  unitary

Prop If  $U_1, U_2$  are unitary matrices, then so is  $U_1 U_2$ .

Pf: Same as for orth.

Thm (Schur Decomposition)

let  $A$  be any  $n \times n$  matrix.

Then there exists a unitary matrix  
 $U$  and upper triangular matrix  $\Delta$

such that

$$A = U\Delta U^+ = U\Delta U^{-1}$$

and the diagonals of  $\Delta$  are the  
eigenvalues of  $A$ .

orthogonal  $\longrightarrow$  unitary

diagonal  $\longrightarrow$  upper triangular

How to compute!

Take  $A$ , compute an eigenvalue  $\lambda, \in \mathbb{C}$

eigenvector  $v, \in \mathbb{C}^n$ .

$$v_1 \rightarrow \frac{v_1}{\|v_1\|} = u_1 \text{ also an eigenvector.}$$

① Find a unitary matrix  $U_1$  w/

$u_1$  as the first column.

$$(u_1 v_2 \dots v_n) \xrightarrow[G \rightarrow]{} (u_1 \tilde{u}_2 \tilde{u}_3 \dots \tilde{u}_n)$$

$$U_1 = (u_1 \tilde{u}_2 \tilde{u}_3 \dots \tilde{u}_n)$$

② Since  $u_1$  is an eigenvector of  $A$  w/ eigenvalue  $\lambda_1$ .

$$\tilde{u}_1^T A u_1 = \begin{pmatrix} \lambda_1 & * & * & * & * \\ 0 & * & * & * & * \\ \vdots & * & * & * & * \\ 0 & * & * & * & * \end{pmatrix}$$

$$U^T A U = \begin{pmatrix} \lambda_1 & * & * & * & * \\ 0 & * & * & * & * \\ \vdots & * & * & * & * \\ 0 & * & * & * & * \end{pmatrix}$$

$$U^T A U = \begin{pmatrix} \lambda_1 & R \\ 0 & C \end{pmatrix}, \text{ this is a block matrix}$$

$\lambda_1$  is  $1 \times 1$  matrix

$R$  is  $1 \times n-1$

$0$  is  $n-1 \times 1$

$C$  is  $n-1 \times n-1$

We're actually done if we can Schur decompose  $C$ .

Assume  $C = V \Gamma V^T$ ,  $V$  is unitary  
 $\Gamma$  upper triangular.

recursive

$$U_2 = \begin{bmatrix} 1 & \vec{v} \\ 0 & V \end{bmatrix}$$

block matrix  
unitary  
 $U^+ = \overline{U}^T$

Claim:

$U_2^+ U_1^+ A \underbrace{U_1 U_2}_U$  is upper  $\Delta$  if eigenvalues on diagonal

$$\underline{U_2^+ (U_1^+ A U_1) U_2} = U_2^+ \begin{pmatrix} \lambda_1 & r \\ 0 & C \end{pmatrix} U_2$$

$$= \underbrace{\begin{bmatrix} 1 & 0 \\ 0 & V^+ \end{bmatrix}}_{\text{red bracket}} \begin{bmatrix} \lambda_1 & r \\ 0 & C \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & V \end{bmatrix}$$

$$= \begin{bmatrix} \lambda_1 & r \\ 0 & V^+ C \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & V \end{bmatrix}$$

$$= \begin{bmatrix} \lambda_1 & rV \\ 0 & V^+ CV \end{bmatrix} = \begin{bmatrix} \lambda_1 & s \\ 0 & \Gamma \end{bmatrix} \quad \text{||}$$

$\Gamma$  is already upper triangular  
w/  $\lambda_2 \dots \lambda_n$  on diagonal!

□

$$\begin{pmatrix} \lambda_1 & & \\ 0 & \ddots & \\ & \ddots & 0 \end{pmatrix} \rightarrow \begin{pmatrix} \lambda_1 & & \\ 0 & \ddots & \\ & \ddots & 0 \end{pmatrix} \quad C$$

diagonalise

$$\rightarrow \begin{pmatrix} \lambda_1 & & & & X \\ 0 & \ddots & & & \times \\ & 0 & \ddots & & \times \\ & & 0 & \ddots & \times \\ & & & 0 & 0 \\ & & & & 0 \end{pmatrix} \quad D$$

$\rightarrow$  go until done.

$$\underline{\text{Ex}} \quad \text{Let } A = \begin{pmatrix} 6 & 4 & -3 \\ -4 & -2 & 2 \\ 4 & 4 & -2 \end{pmatrix}$$

Not a diagonalizable matrix

$$P_A(\lambda) = 2\lambda^2 - \lambda^3 = 0$$

$$= \lambda^2(2-\lambda) = 0$$

$$\underbrace{\lambda=0, \lambda=0}_{\text{alg mult.}} , \lambda=2$$

$$\begin{aligned} &\text{alg mult.} \\ &= 2 \end{aligned}$$

$$V_{\lambda=0} = \ker(A - 0I) = \ker(A)$$

$$= \text{span} \left( \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix} \right) \quad \frac{I-D}{\text{not diagonalizable}} < 2$$

row  
reduce

$$V_{\lambda=2} = \ker(A - 2I) = \text{span} \left( \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \right)$$

Let's compute Schur decomposition to A

(let's pick  $\lambda_1 = 2$ ,  $v_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ )  
as unit eigenvector.

① Find a unitary matrix w/  
 $\frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$  as the first column.

$$\frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$$\xrightarrow{\text{G-S}} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$$

$$U_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 & 0 \\ 1 & -1 & 0 \\ 0 & 0 & \sqrt{2} \end{pmatrix}$$

diagonalizes first  
column only

$$\overline{U}_1^T A U_1 = \begin{pmatrix} 2 & -9 & \frac{5\sqrt{2}}{2} \\ 0 & 2 & -\frac{1}{\sqrt{2}} \\ 0 & 4\sqrt{2} & -2 \end{pmatrix}$$

Take this  
and this  
in the  
2x2  
case.

diagonalize this part

$$C = \begin{pmatrix} 2 & \frac{1}{\sqrt{2}} \\ 4\sqrt{2} & -2 \end{pmatrix} \rightsquigarrow \boxed{\lambda = 0} \quad \boxed{\lambda = 0}$$

But  $V_0 = \ker(A - 0I) = \ker(A)$   
 $= \text{span}\left(\begin{pmatrix} 1 \\ 2\sqrt{2} \end{pmatrix}\right)$

Unit vector is  $u_2 = \frac{1}{3} \begin{pmatrix} 1, 2\sqrt{2} \end{pmatrix}$

$$u_2, e_2 \xrightarrow{G-S} \frac{1}{3} \begin{pmatrix} 1 \\ 2\sqrt{2} \end{pmatrix}, \frac{1}{3} \begin{pmatrix} 2\sqrt{2} \\ -1 \end{pmatrix}$$

$$V = \boxed{\frac{1}{3} \begin{pmatrix} 1 & 2\sqrt{2} \\ 2\sqrt{2} & -1 \end{pmatrix}}$$

$$VCV^T = \begin{pmatrix} 0 & \frac{1}{\sqrt{2}} \\ 0 & 0 \end{pmatrix}$$

→ Plug

into  
previous  
step

$$U_2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \sqrt{3} & 2\sqrt{3} \\ 0 & 2\sqrt{3} & -1/3 \end{pmatrix}$$



$$U_2^+ U_1^+ A U_1 U_2 = \Delta$$

$$(U_1 U_2)^T A \underbrace{U_1 U_2}_{U} = \Delta$$

$\underbrace{\quad \quad \quad}_{U}$

upper  $\Delta$  w/  $\lambda$  or  
diagonal

$$U = \frac{1}{\sqrt{2}} \begin{pmatrix} -1 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & \sqrt{2} \end{pmatrix} \frac{1}{\sqrt{3}} \begin{pmatrix} 3 & 0 & 0 \\ 0 & 1 & 2\sqrt{2} \\ 0 & 0 & -1 \end{pmatrix}$$

$$= \begin{pmatrix} \frac{-1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{2}{\sqrt{3}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \frac{2}{\sqrt{3}} \\ 0 & \frac{2\sqrt{2}}{3} & \frac{-1}{\sqrt{3}} \end{pmatrix}$$

$$A = \underline{U} \underline{\Delta} \overline{U^T}$$

$$\Delta = U_2^+ \underbrace{(U_1^+ A U_1)}_{\text{matrix}} U_2$$

$$= U_2^+ \begin{pmatrix} 2 & -8 & \frac{5\sqrt{2}}{2} \\ 0 & 2 & -\frac{1}{\sqrt{2}} \\ 0 & 4\sqrt{2} & \end{pmatrix} U_2^+$$

$$= \begin{pmatrix} 2 & \gamma_3 & -\frac{3\gamma}{3\sqrt{2}} \\ 0 & 0 & \gamma/\sqrt{2} \\ 0 & 0 & 0 \end{pmatrix}$$

eigenvalues  
or  
diagonal

$$A = U \Delta U^{-1}$$

$$U = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{3\sqrt{2}} & \frac{2}{3} \\ \frac{1}{\sqrt{2}} & \frac{1}{2\sqrt{2}} & \frac{2}{3} \\ 0 & \frac{2\sqrt{2}}{3} & \frac{1}{3} \end{pmatrix}$$

# Jordan Canonical form

Thm Let  $A$  be a matrix,  $n \times n$ .

then  $A = SJS^{-1}$  where  $J$

is a matrix of the form

$$\left[ \begin{array}{c} J_{\lambda_1, k_1} \\ \vdots \\ J_{\lambda_n, k_n} \end{array} \right] .$$

block matrix

What is  $J_{\lambda, k}$  in general?

$J_{\lambda,k}$  is called a Jordan block.

a  $k \times k$  matrix w/  $\lambda$  on the diagonal and 1's on the superdiagonal (diagonal above).

$$\begin{bmatrix} \lambda & 1 & & \\ & \lambda & 1 & \\ & & \ddots & \\ & & & \ddots & 1 \end{bmatrix}$$

$$k \times k$$
$$J_{5,3} = \begin{bmatrix} 5 & 1 & \\ & 5 & 1 \\ & & 5 \end{bmatrix} \quad J_{3,1} = [3]$$

$$J_{-4,4} = \begin{bmatrix} -4 & 1 & & \\ & -4 & 1 & \\ & & -4 & 1 \\ & & & -4 \end{bmatrix}$$

$$A = SJS^{-1}$$

$$J = \begin{bmatrix} & J_{2,1} \\ J_{2,2} & \\ J_{3,2} & \end{bmatrix}$$

The matrix  $A$  is shown as a block diagonal matrix with three blocks:

- Block 1 (top-left):  $\begin{pmatrix} 2 & & \\ & 1 & \\ & & 2 \end{pmatrix}$
- Block 2 (middle):  $\begin{pmatrix} 3 & 1 & \\ & 3 & \\ & & 3 \end{pmatrix}$

The super-diagonal element  $1$  is highlighted in red. Red arrows point from the text "except for these 1's." to this element.

**But  $\lambda = 2$  only has one eigenvector**

**$\lambda = 2$   $\lambda = 2$   $\lambda = 2$**

**$\lambda = 3$   $\lambda = 3$   $\lambda = 3$**

**lack of an eigenvector**

**almost diagonalization**

**except for these 1's.**

**A 1 in the superdiagonal corresponds to a lack of an eigenvector**

Let's take  $A = \begin{pmatrix} 6 & 4 & -3 \\ -4 & 2 & 2 \\ 4 & 4 & -2 \end{pmatrix}$

Spoiler ...

$$A = SJS^{-1}$$

$$\lambda = 0, \lambda = 0$$

$$\lambda = 2$$

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

We  
S<sup>-1</sup> need one  
more  
eigenvector  
for  $\lambda = 0$ .

$$V_0 = \text{span} \left( \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix} \right) \text{ despite } \begin{matrix} \lambda = 0 \\ \lambda = 0 \end{matrix}$$

Added a "kind of eigenvector"  
which added a 1 above  
the diagonal.

Normally  $A = \begin{pmatrix} 6 & 4 & -3 \\ -4 & 2 & 2 \\ 4 & 4 & -2 \end{pmatrix}$

 $A = SJS^{-1}, \quad S = \begin{pmatrix} v_1 & v_2 & v_3 \end{pmatrix}$

eigenvectors as columns

$\lambda = 0, \lambda = 0$

$v = \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix}$

$\lambda = 2$

$\begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$

$S = \begin{pmatrix} 1 & ?? & -1 \\ 0 & v_2 & 1 \\ 2 & ?? & 0 \end{pmatrix}$

$v_2$  is a generalized eigenvector

$A = SJS^{-1}$

$AS = SJ$

$$A = \begin{pmatrix} 6 & 4 & -3 \\ -4 & -2 & 2 \\ 4 & 4 & -2 \end{pmatrix}$$

$$\begin{pmatrix} 6 & 4 & -3 \\ -4 & -2 & 2 \\ 4 & 4 & -2 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 0 & v_2 \\ 2 & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & -1 \\ 0 & v_2 \\ 2 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 0 & 2 \end{pmatrix}$$

$$(A \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix}, Av_2, A \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix})$$

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

$$\begin{pmatrix} 0 & Av_2 & -2 \\ 0 & 0 & 2 \end{pmatrix} = \begin{pmatrix} 0 & 1 & -2 \\ 0 & 0 & 2 \end{pmatrix}$$

$$\underline{Av_2 = \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix}} \quad v_2 = \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}$$

Given a linear transformation

$$T: V \rightarrow W, \text{ vector space}$$

$$\langle v, \tilde{v} \rangle_1 \quad \langle w, \tilde{w} \rangle_2$$

The adjoint  $T^*: W \rightarrow V$  is

$$\langle T(v), w \rangle_2 = \langle v, T^*(w) \rangle_1$$

such that

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Formula if  $A: \mathbb{R}^n \rightarrow \mathbb{R}^m$

$$x^T K x \quad y^T L y$$

$K, L$  pos. def.

$$A^* = K^{-1} A^T L$$

$$n \times m \quad n \times n \quad n \times m$$

If dot product, then  $A^* = A^T$

In F.S.I  
 $K = L$

$$R_\theta = \begin{pmatrix} \cos\theta - \lambda & -\sin\theta \\ \sin\theta & \cos\theta - \lambda \end{pmatrix}$$

$$\cos^2\theta - 2\cos\theta\lambda + \lambda^2 + \sin^2\theta$$

$$= \lambda^2 - 2\cos\theta\lambda + 1$$

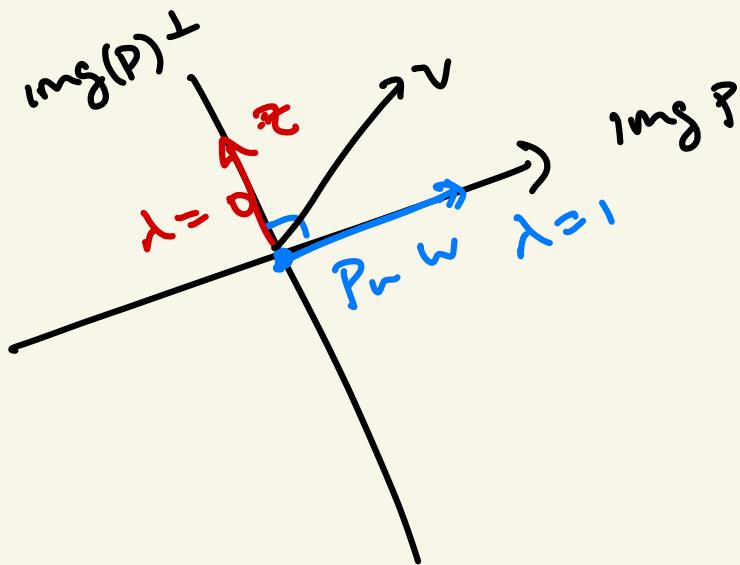
$$\lambda = \frac{\pm \sqrt{ }}{ }$$

$$(mg(P)^\perp = \text{ker}(P)$$

$$v \in \text{Im}(P)^\perp \Rightarrow Pv = 0.$$

If  $Pv$ ,  $v \in \text{Im}(P)^\perp$ .

then  $Pv \in \text{Im}(v)$



let  $v = w + z$   $w \in \text{img } P$   
 $z \in \text{img } P^\perp$

$$Pv = Pv + Pz$$

$$= Pw + Pz$$

$$Pv = Pw + Pz$$

$$v - w - z \in \text{ker } P$$

$$w + z - v - z \in \text{ker } P$$

$$w - v \in \text{ker } P$$

$$w - v \cdot P$$

$$Pw - v \in \text{ker } P$$

$$P_V \cdot (v - P_V v) = P_V \cdot v - P_V \cdot P_V v$$

$$= v^T P_V^T - v^T P^T P_V = 0$$

$$v^T (P^T - P^T P) v = 0$$

$$A^+ = K^{-1} A K$$

$$(v_1, v_2) = e_i$$

$$2v_1w_1 + 3v_2w_2 \quad \begin{pmatrix} w_1 \\ w_2 \end{pmatrix} = e_j$$

$$= (v_1, v_2) K \begin{pmatrix} w_1 \\ w_2 \end{pmatrix} \quad e_i^T K e_j = k_{ij}$$

$$K = \frac{w_1}{w_2} \begin{pmatrix} v_1 & v_2 \\ 2 & 0 \\ 0 & 3 \end{pmatrix} \quad !!$$

$$K = \begin{pmatrix} 2 & 3 \\ 0 & 3 \end{pmatrix}$$

$$\begin{aligned}
 \underline{Aw} &= Acv = cAv \\
 &= c\lambda v \\
 &= \lambda(cv) \\
 &= \underline{\lambda w}
 \end{aligned}$$

$w = cv$  is an eigenvector

even though  $c \in \mathbb{C}$ .

$$\begin{aligned}
 \det(U - \lambda I) &= \det \begin{pmatrix} u_{11} - \lambda & * & * & * \\ * & u_{22} - \lambda & * & * \\ * & * & \ddots & * \\ * & * & * & u_{nn} - \lambda \end{pmatrix} \\
 &= (u_{11} - \lambda)(u_{22} - \lambda) \dots (u_{nn} - \lambda)
 \end{aligned}$$

Ux 8.2.20

$$A^2 \rightarrow \lambda^2$$

$$P^2 - \lambda^2 I = P - \lambda^2 I \neq 0$$

$V_{\lambda^2} \cup V_\lambda$  compare these.

$$\lambda^2 v = P^2 v = Pv = \lambda v$$

$v \neq 0$

$$\begin{aligned}\lambda^2 &= \lambda \\ \lambda^2 - \lambda &= \lambda(\lambda - 1) \\ \Rightarrow \lambda &= 0, 1\end{aligned}$$