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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.

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

$\Lambda$ , diagonal matrix  
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

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

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

$$q(x) = (Q^T x)^T \Lambda (Q^T x)$$

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

$$\Rightarrow x = Qy$$

$$q(y) = y^T \Lambda y$$

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

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

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

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

$\Leftarrow$  If  $\lambda_i > 0$

$$q(x) = y^T \Lambda y = \sum \lambda_i y_i^2$$

Since  $\lambda_i > 0$ ,  $q(x) = \sum \lambda_i y_i^2 > 0$

□



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

$$\lambda = 2, \quad \lambda = 4$$

$$v = \begin{pmatrix} -1 \\ 1 \end{pmatrix} \quad v = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\begin{array}{ccc} G-S & \downarrow & G-S \\ \frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 1 \end{pmatrix} & & \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \end{array}$$

$$4, 2 > 0$$

so

A is positive definite!

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

Spectral decomposition

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

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

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

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

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

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

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

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

on the other hand

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

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

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

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

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

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

$$y_2 = -x_1 + x_2$$

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

$$= 3x_1^2 + 2x_1x_2 + 3x_2^2 > 0$$

Recall that

$K$  being pos. def

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

defined an inner product

$$f(x, y) \quad \text{min/max} \quad D^2 f = 0$$

$$Hf = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_1 \partial x_2} & \frac{\partial^2 f}{\partial x_2^2} \end{bmatrix} \quad \text{if } Hf \text{ pos def} \Rightarrow \text{min value}$$

Ex Consider

$$q(x, y, z) = x^2 + 2xz + y^2 - 2yz > 0 \quad ?$$

write  $q(x, y, z) = (x \ y \ z) K \begin{pmatrix} x \\ y \\ z \end{pmatrix}$

and find the eigenvalues of  $K$ !

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

Compute eigenvalues  $\rightarrow$

$$\boxed{\lambda = -1}$$

not positive!

$$\lambda = 1$$

$$\lambda = 2$$

So  $q(x) \not> 0$  for all  $x$ .

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

eigenvector

Thm Let  $K$  be a symmetric matrix

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

(Ex  $\begin{array}{ccc|c} 1 & 0 & 1 & \\ \hline 0 & 1 & -1 & \\ 1 & -1 & 0 & \end{array} \quad K_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ )

Then TFAE

1)  $K$  is positive definite

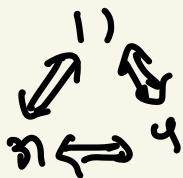
2) all eigenvalues of  $K > 0$

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  $U$ .

$$U^H = \overline{U^T} \text{ OR } U^\dagger = \overline{U^T} = \bar{u}^T$$

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  
of  $\mathbb{C}^n$ .



Pf: If  $U$  is unitary, then

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

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

But  $U = (u_1 \ u_2 \ \dots \ u_n)$

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

$$U^{-1}U = \begin{pmatrix} \overline{u_1} \\ \overline{u_2} \\ \vdots \\ \overline{u_n} \end{pmatrix} \begin{pmatrix} u_1 & \dots & u_n \end{pmatrix} = I$$

*need*  $\rightarrow$

$$\Leftrightarrow \begin{cases} \overline{u_i} \cdot u_j = 0 & \text{if } i \neq j \\ \overline{u_i} \cdot u_i = 1 & \text{if } i = j \end{cases}$$

$\Leftrightarrow \{u_i\}_i$  form an orthonormal basis of  $\mathbb{C}^n$ !

□

$$\underline{\text{Ex}} \quad u = \begin{bmatrix} i & 0 \\ 0 & i \end{bmatrix} = (ie_1, ie_2)$$

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

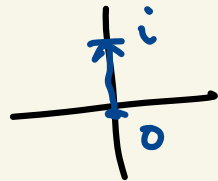
$$\overline{u^T} u = \begin{bmatrix} -i & 0 \\ 0 & -i \end{bmatrix} \begin{bmatrix} i & 0 \\ 0 & i \end{bmatrix}$$

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

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

$ie_1, ie_2$  should be orthonormal

$$\begin{aligned} \|ie_1\| &= |i| \|e_1\| \\ &= 1 \cdot 1 = 1 \end{aligned}$$



If we want to generalize the spectral decomposition

$$A = \underbrace{Q}_{\text{orth.}} \Lambda Q^T$$

to a general matrix.

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

orth.  $\dagger \mathbb{C}^n \rightarrow$  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

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

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  $\lambda_1 \in \mathbb{C}$

eigenvector  $v_1 \in \mathbb{C}^n$

$$v_1 \longrightarrow \frac{v_1}{\|v_1\|} = u_1 \quad \text{also an eigen vector.}$$

① Find a unitary matrix  $U_1$  w/

$u_1$  as the first column.

$$(u_1 \ v_2 \ \dots \ v_n) \xrightarrow{G-S} (u_1 \ \tilde{u}_2 \ \tilde{u}_3 \ \dots \ \tilde{u}_n)$$

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

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

$$U_1^{-1} A U_1 = \begin{pmatrix} \lambda_1 & * & * & * & * \\ 0 & * & * & * & * \\ \vdots & * & * & * & * \\ 0 & & & & \ddots \end{pmatrix}$$

$$U_1^{-1} A U_1 = \begin{pmatrix} \lambda_1 & * & * & * & * \\ 0 & * & * & * & * \\ \vdots & * & * & * & * \\ 0 & & & & \ddots \end{pmatrix}$$

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

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

$\Gamma$  is  $1 \times n-1$

$\vec{0}$  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

$$U_2 = \begin{bmatrix} 1 & \vec{0} \\ \vec{0} & V \end{bmatrix} \quad \begin{array}{l} \text{block matrix} \\ \text{unitary} \\ U^\dagger = \overline{U}^T \end{array}$$

Claim!

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

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

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

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

$$= \begin{bmatrix} \lambda_1 & rV \\ 0 & V^\dagger C V \end{bmatrix} = \begin{bmatrix} \lambda_1 & s \\ 0 & \Gamma \end{bmatrix} \quad \Delta \parallel$$

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

$$\begin{pmatrix} \lambda_1 & & & & \\ & \ddots & & & \\ & & \lambda_1 & & \\ & & & \ddots & \\ & & & & 0 \end{pmatrix}$$

diagonalizable

$$\begin{pmatrix} \lambda_1 & & & & \\ & \ddots & & & \\ & & \lambda_1 & & \\ & & & \ddots & \\ & & & & c \end{pmatrix}$$

An orange box highlights the diagonal elements  $\lambda_1, \dots, \lambda_1, c$ . A red arrow points from the  $c$  element down to the next matrix.

$$\begin{pmatrix} \lambda_1 & \times & \times & \times & \times \\ & \lambda_2 & \times & \times & \times \\ & & 0 & & \\ & & 0 & & \\ & & 0 & & \\ & & 0 & & \\ & & 0 & & \end{pmatrix}$$

A horizontal line is drawn above the bottom row of zeros. An orange box highlights the diagonal elements  $\lambda_1, \lambda_2, 0, 0, 0$ . A red 'x' is above the first zero, and a red circle with an 'x' is around the  $\lambda_1$  in the top-left position.

→ go until done.



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

Not a diagonalizable matrix

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

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

$$\lambda = 0, \lambda = 0, \lambda = 2$$

alg mult.  
= 2

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

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

row  
reduce

$$\underline{1-D} < 2$$

not diagonalizable

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

Let's compute Schur decomposition of  $A$

Let's pick  $\lambda_1 = 2$ ,  $v_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}$

|| unit eigenvector.

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

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

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

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

diagonalizes first column only

$$\overline{U_1}^T A U_1 = \begin{pmatrix} 2 & -8 & \frac{1}{\sqrt{2}} \\ 0 & 2 & \frac{1}{\sqrt{2}} \\ 0 & 4\sqrt{2} & 2 \end{pmatrix}$$

diagonalize this part

Take this and do this in the 2x2 case.

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

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

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

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

$\frac{1}{3} \begin{pmatrix} 1 \\ 2\sqrt{2} \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix}$

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

$$V C V^T = \begin{pmatrix} 0 & \sqrt{2} \\ 0 & 0 \end{pmatrix}$$

plug  
into  
previous  
step

$$u_2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1/3 & 2\sqrt{2}/3 \\ 0 & 2\sqrt{2}/3 & -1/3 \end{pmatrix}$$

$$u_2^T u_1^T A u_1 u_2 = \Delta$$

$$(u_1 u_2)^T A \underbrace{u_1 u_2}_u = \Delta$$

$$\underbrace{\hspace{10em}}_u$$

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

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

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

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

$$\Delta = u_2^+ \underbrace{(u_1^+ A u_1)}_{\text{scalar}} u_2$$

$$= u_2^+ \begin{pmatrix} 2 & -8 & \frac{1}{\sqrt{2}} \\ 0 & 2 & \frac{1}{\sqrt{2}} \\ 0 & 4\sqrt{2} & 1 \end{pmatrix} u_2^+$$

$$= \begin{pmatrix} 2 & \frac{2}{3} & \frac{-37}{3\sqrt{2}} \\ 0 & 0 & \frac{9}{\sqrt{2}} \\ 0 & 0 & 0 \end{pmatrix}$$

eigenvalues  
or  
diagonal

$$A = U \Delta U^T$$

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

# Jordan Canonical form

Then let  $A$  be a matrix,  $n \times n$ .

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

is a matrix of the form

$$\left[ \begin{array}{c} \underbrace{J_{\lambda_1, k_1}} \\ \underbrace{J_{\lambda_2, k_2}} \\ \vdots \\ \underbrace{J_{\lambda_n, k_n}} \end{array} \right] \text{ block matrix}$$

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).

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

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

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

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

$$J = \begin{bmatrix} \lambda_1 & & & & & \\ & \lambda_1 & & & & \\ & & \lambda_1 & & & \\ & & & \lambda_2 & & \\ & & & & \lambda_2 & \\ & & & & & \ddots \\ & & & & & & \lambda_n \end{bmatrix}$$

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

$$= \begin{bmatrix} \boxed{2} & & \\ & \begin{bmatrix} 2 & 1 \\ & 2 \end{bmatrix} & \\ & & \begin{bmatrix} 3 & 1 \\ & 3 \end{bmatrix} \end{bmatrix}$$

But  $\lambda = 2$   
only lacked  
1 eigen  
vector

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

$\lambda = 3$   
 $\lambda = 3$   
lacked 1

Almost diagonalizable

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 ...

$A = S J S^{-1}$   $\lambda = 0, \lambda = 0$   
 $\lambda = 2$

$= S \begin{pmatrix} \boxed{0} & \boxed{1} & 0 \\ 0 & 0 & 0 \\ 0 & 0 & \boxed{2} \end{pmatrix} S^{-1}$  we need one more eigenvector for  $\lambda = 0$ .

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

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

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

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

$$\lambda = 2$$

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

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

$v_2$  is a generalized eigenvector

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

$$AS = SJ$$

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

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

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

$$\left( A \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix} \quad Av_2 \quad A \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} \right)$$

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

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

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

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Given a linear transformation

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

$$\langle v, \tilde{v} \rangle_1$$

$$\langle w, \tilde{w} \rangle_2$$

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

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

Such that

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

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

$K, L$  pos. def.

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

$n \times m$

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

In 7.5.1  
 $K=L$

If

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

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

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

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

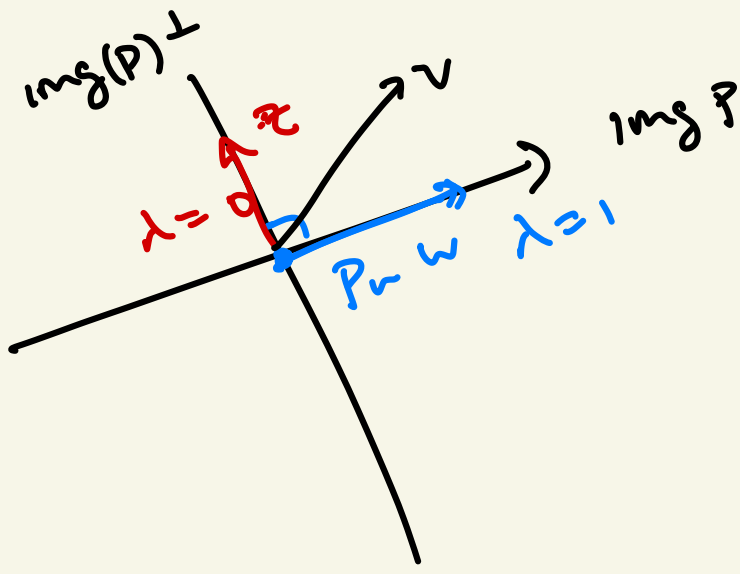
$$\lambda = \frac{\pm \sqrt{\quad}}{\quad}$$

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

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

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

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



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

$Pv = Pw + Pz$   $w - w' \in \ker P$   
 $= Pw' + Pz$   $Pw' - w' \in \ker P$   
 $Pv = Pw' + Pz$

$v - w' - z \in \ker P$

$w + z - w' - z \in \ker P$   
 $w - w' \in \ker P$

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

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

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


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

$$(v_1, v_2) = e_i$$

$$2v_1w_1 + 3v_2w_2$$

$$\begin{pmatrix} w_1 \\ w_2 \end{pmatrix} = e_j$$

$$= (v_1, v_2) K \begin{pmatrix} w_1 \\ w_2 \end{pmatrix}$$

$$e_i^T K e_j = k_{ij}$$

" "

$$K = \frac{1}{w_2} \begin{pmatrix} v_1 & v_2 \\ 2 & 0 \\ 0 & 3 \end{pmatrix}$$

$$K_2 = \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix}$$

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

$w = cv$  is an eigenvector  
even though  $c \in \mathbb{C}$ .

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



Ux 8.2.20

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

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

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

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

$$v \neq 0$$

$$\lambda^2 = \lambda$$

$$\lambda^2 - \lambda = \lambda(\lambda - 1)$$

$$\Rightarrow \lambda = 0, 1$$