

an  $n \times n$  matrix with **complex** entries has  **$n$**  complex eigenvalues

an  $n \times n$  matrix with **real** entries has  **$n$**  complex eigenvalues,

an  $n \times n$  **symmetric** matrix with **real** entries has  **$n$**  real eigenvalues and  **$n$**  orthonormal eigenvectors

acceptable eigenvalues for a real  $3 \times 3$

•  $\lambda_1 = 2$ ,  $\lambda_2 = \frac{7}{2}$ ,  $\lambda_3 = \pi$  ✓

•  $\lambda_1 = 7$ ,  $\lambda_2 = 5 + 4i$ ,  $\lambda_3 = 5 - 4i$  ✓

•  $\lambda_1 = 3 + 4i$ ,  $\lambda_2 = 3 - 4i$ ,  $\lambda_3 = 7 + 5i$  ✗

but the non-real ones among the eigenvalues come in complex conjugate pairs (also the respective eigenvectors will be complex conjugate)

e.g.  $\lambda_1 = 3$  and  $\lambda_2 = -7$  ✓  
 $\begin{bmatrix} 3 & 0 \\ 0 & -7 \end{bmatrix}$

e.g.  $\lambda_1 = i$  and  $\lambda_2 = -i$  ✓  
 $\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$

e.g.  $\lambda_1 = 3$  and  $\lambda_2 = 7 - 5i$  ✗

e.g.  $\lambda_1 = 3 + 2i$  and  $\lambda_2 = 1 - 8i$  ✗

2x2 symmetric

$$\begin{bmatrix} a & b \\ b & c \end{bmatrix}$$

$$\rightsquigarrow p(\lambda) = \lambda^2 - \lambda(a+c) + ac - b^2$$

$$\lambda_{1,2} = \frac{a+c \pm \sqrt{(a+c)^2 - 4ac + 4b^2}}{2}$$

$$= \frac{a+c \pm \sqrt{(a-c)^2 + 4b^2}}{2}$$

$\rightsquigarrow > 0$  } real

• Why are the eigenvectors of symmetric matrices mutually orthogonal?

•  $v$  is an eigenvector of  $S$  corresp. to  $\lambda \Rightarrow Sv = \lambda v$

•  $v'$  is an eigenvector of  $S$   $\rightarrow$  sym. matrix  
corresp. to  $\lambda' \Rightarrow Sv' = \lambda' v'$

assume  $\boxed{\lambda \neq \lambda'}$

$$\lambda \cdot v^T v' = (Sv)^T v' \quad v'^T (Sv') = \lambda' v'^T v' \Rightarrow v^T v' = 0 \Rightarrow v \perp v'$$

$$\overset{\parallel}{v^T} \overset{\parallel}{S^T} v' = \overset{\parallel}{v^T} \overset{\parallel}{S} v'$$

$\hookrightarrow$  b/c  $S$  is symmetric

b/c  $\lambda \neq \lambda'$

$S \leadsto$  an  $n \times n$  symmetric matrix

has  $n$  mutually orthonormal eigenvectors  $\{q_1, \dots, q_n\}$  will give an orthonormal basis of  $\mathbb{R}^n$

$$Q = [q_1 | \dots | q_n]$$

corresponding  
to eigenvalues

$$\lambda_1 \dots \lambda_n$$

because the columns of  $Q$  are orthonormal,  $Q$  is an orthogonal matrix

$$Q^{-1} = Q^T$$

Diagonalization theorem:

$$S = Q \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} Q^{-1}$$

$$= \underbrace{Q}_{\text{rotation}} \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \underbrace{Q^T}_{\text{inverse rotation}}$$

scaling

2x2 case,

$$Q = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$



$$S = Q \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} Q^T$$

$$= [q_1 | \dots | q_n] \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \begin{bmatrix} q_1^T \\ \vdots \\ q_n^T \end{bmatrix} = \lambda_1 \underbrace{q_1 q_1^T}_{n \times n} + \dots + \lambda_n \underbrace{q_n q_n^T}_{n \times n}$$

rank 1  $n \times n$  matrix

given  $v \in \mathbb{R}^n$ , how do I compute  $Sv \in \mathbb{R}^n$

(Singular Value Decomposition  
..... in a couple of weeks)

$$v = c_1 q_1 + \dots + c_n q_n$$

!! computationally efficient

$$Sv = \underline{c_1 \lambda_1 q_1 + \dots + c_n \lambda_n q_n}$$

Def: if a symmetric real matrix  $S$  has all its eigenvalues

- positive / negative,  $S$  is called positive / negative definite
- nonnegative / nonpositive,  $S$  is called semidefinite

$\lambda_i > 0$

$\lambda_i \geq 0$

any symmetric real matrix has as many positive eigenvalues as positive pivots

Energy (of a vector  $v$  in relation to a symmetric matrix  $S$ )

$v^T S v$

Theorem

$S$  is positive definite  $\implies$  energy of any  $v \neq 0$  is  $> 0$

$S$  is positive semidefinite  $\implies$  energy of any  $v$  is  $\geq 0$

any  $S = A^T A$  is positive semidefinite

$$v^T S v = v^T A^T A v = (Av)^T Av = \|Av\|^2 \geq 0$$

$S = A^T A$  positive definite if

$Av \neq 0$  for any nonzero  $v$   
i.e. the columns of  $A$  are independent

Proof:

if  $v = c_1 g_1 + \dots + c_n g_n$ , then  $Sv = c_1 \lambda_1 g_1 + \dots + c_n \lambda_n g_n$

$$v^T S v = \sum_{1 \leq i, j \leq n} c_i \lambda_j c_j g_i^T g_j$$

$\rightarrow = 0$  unless  $i=j$   
also  $g_i^T g_i = 1$

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

if  $\lambda_1, \dots, \lambda_n > 0$

energy

$$S = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$$

$$\lambda_{1,2} = \frac{a+c \pm \sqrt{(a-c)^2 + 4b^2}}{2}$$

easy to prove that

$\lambda_1$  and  $\lambda_2$  are  $> 0$

$\Downarrow$

$\underbrace{T_n S}_{a+c}$  and  $\underbrace{\det S}_{ac-b^2} > 0$

$$v = \begin{bmatrix} x \\ y \end{bmatrix}$$

$$v^T S v = \begin{bmatrix} x & y \end{bmatrix} \begin{bmatrix} a & b \\ b & c \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = ax^2 + 2bxy + cy^2$$

$\downarrow$   
equation of a  
conic (ellipse, hyperbola)  
curve in  $\mathbb{R}^2$