

# ① Spectral Theorem 🤪

Given  $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$  linear transformation

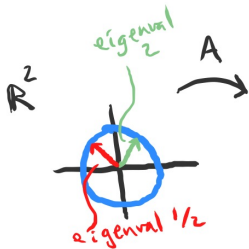
Nice: There is a basis for  $\mathbb{R}^n$  of eigenvectors for  $T$

Really nice: There is an orthonormal basis for  $\mathbb{R}^n$  of eigenvectors for  $T$

Spectral Theorem: If  $A$  is an  $n \times n$  symmetric matrix whose entries are real numbers then

$$A^T = A$$

- ① There is an orthonormal basis of  $\mathbb{R}^n$  consisting of eigenvectors for  $T$
- ② All eigenvalues of  $A$  are real



① Which of these are symmetric?

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

Not symmetric

$$\begin{bmatrix} 1 & -1 \\ -1 & 3 \end{bmatrix}$$

symmetric

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

Not symmetric

$$\begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}^T = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix}$$

$$\begin{bmatrix} 1 & -1 \\ -1 & 3 \end{bmatrix}^T = \begin{bmatrix} 1 & -1 \\ -1 & 3 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}^T = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

② Suppose  $A$   $n \times n$  has eigenvectors  $\vec{v}_1, \dots, \vec{v}_n$  which are orthonormal and have eigenvals  $\lambda_1, \dots, \lambda_n$

$$\text{Show } A = \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix} \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix}^T$$

$$\text{and } A^T = A \quad \text{"P" \quad "D"}$$

By diagonalization:  $A = \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix} \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix}^{-1}$

Since  $\begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix}$  is orthogonal, & square  $\begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix}^{-1} = \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix}^T$

$$\begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix}^T \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix} = \begin{bmatrix} - & \vec{v}_1 & - \\ & \vdots & \\ - & \vec{v}_n & - \end{bmatrix} \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix} = \begin{bmatrix} \vec{v}_1 \cdot \vec{v}_1 & \vec{v}_1 \cdot \vec{v}_2 & \dots \\ \vec{v}_2 \cdot \vec{v}_1 & \vec{v}_2 \cdot \vec{v}_2 & \dots \\ \vdots & \vdots & \ddots \\ \vec{v}_n \cdot \vec{v}_1 & \vec{v}_n \cdot \vec{v}_2 & \dots \end{bmatrix}$$

$$A^T = (PDP^T)^T = (P^T)^T D^T P^T = PDP^T = A = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$$

# Singular Value Decomposition

$$(A = U \Sigma V^T) \begin{matrix} \rightarrow \text{orthogonal} \\ \rightarrow \text{diagonal} \\ \rightarrow \text{orthogonal} \end{matrix}$$

Idea:  $A^T A$  is always symmetric

$\Rightarrow A^T A$  has an orthonormal basis of eigenvectors & its eigenvals are all real

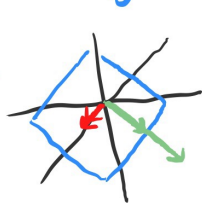
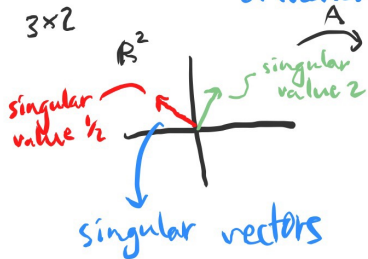
Thm The eigenvals of  $A^T A$  are nonnegative numbers

$A^T A$  eigenvals  $\sigma_1^2 \geq \sigma_2^2 \geq \dots \geq \sigma_n^2 \geq 0$

← singular values of  $A$

orthonormal eigenvectors  $\vec{v}_1, \dots, \vec{v}_n$

$A$   $3 \times 2$



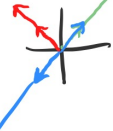
$\mathbb{R}^3$

↑ right singular vectors

Singular vector w/ largest singular value tells you direction that  $A$  expands the most

③ Suppose  $A$  is a  $2 \times 2$  matrix and  $\vec{v}_1, \vec{v}_2$  are eigenvectors for  $A$  with eigenvals  $3, 2$  and they are orthonormal. Find the maximum value of  $\|A\vec{v}\|$  over all unit vectors  $\vec{v}$ . 3

Hint: try writing  $\vec{v}$  as a linear combination of  $\vec{v}_1$  &  $\vec{v}_2$  & calculating  $\|\vec{v}\|^2$  &  $\|A\vec{v}\|^2$



Let  $\vec{v} \in \mathbb{R}^2$ . Because  $\vec{v}_1, \vec{v}_2$  form a basis for  $\mathbb{R}^2$ , there are  $a, b \in \mathbb{R}$  s.t.  $\vec{v} = a\vec{v}_1 + b\vec{v}_2$

$$\begin{aligned} \|\vec{v}\|^2 &= \vec{v} \cdot \vec{v} = (a\vec{v}_1 + b\vec{v}_2) \cdot (a\vec{v}_1 + b\vec{v}_2) = a^2\vec{v}_1 \cdot \vec{v}_1 + ab\vec{v}_1 \cdot \vec{v}_2 \\ &\quad + ba\vec{v}_2 \cdot \vec{v}_1 + b^2\vec{v}_2 \cdot \vec{v}_2 \\ &= a^2 + b^2 \end{aligned}$$

$$\begin{aligned} \|A\vec{v}\|^2 &= \|A(a\vec{v}_1 + b\vec{v}_2)\|^2 = \|a \cdot A\vec{v}_1 + b \cdot A\vec{v}_2\|^2 = \|3a\vec{v}_1 + 2b\vec{v}_2\|^2 \\ &= (3a)^2 + (2b)^2 = 9a^2 + 4b^2 \end{aligned}$$

largest when  
assuming  $a^2 + b^2 = 1$   $a^2 = 1$   $b^2 = 0$

If  $\vec{v}_1, \dots, \vec{v}_n$  is a basis of orthonormal eigenvectors of  $A$  with real eigenvalues then the direction in which  $A$  expands the most is the eigenvector w/ largest eigenvalue

