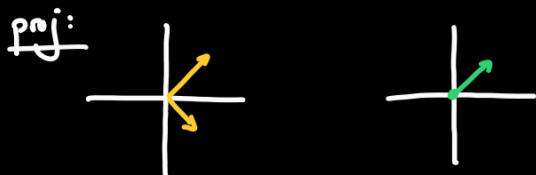


Claim: For every linear transformation we have a basis such that:

$$A \quad n \times m \\ \mathbb{R}^m \rightarrow \mathbb{R}^n$$

1. The basis is orthogonal.

2. The image of the basis is orthogonal.



Example: Consider the linear transformation given by  $A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$ .

orthonormal

(a) Find a basis  $\mathcal{B} = \{\vec{v}_1, \vec{v}_2\}$  of  $\mathbb{R}^2$  such that  $L(\vec{v}_1)$  and  $L(\vec{v}_2)$  are orthogonal.

Hint: This is an application of the Spectral Theorem.

We need a symmetric matrix.  $A^T A$

$$A^T A = S D S^T$$

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

$$\ker(A^T A - \lambda I_2)$$

$$\lambda_1 = \frac{3}{2} + \frac{\sqrt{5}}{2}, \quad \lambda_2 = \frac{3}{2} - \frac{\sqrt{5}}{2}$$

$$A^T A \vec{x} = \lambda \cdot \vec{x}$$

$$\vec{v}_1 = \begin{bmatrix} -\frac{1}{2} + \frac{\sqrt{5}}{2} \\ \frac{\sqrt{5}}{2} \end{bmatrix}, \quad \vec{v}_2 = \begin{bmatrix} -\frac{1}{2} - \frac{\sqrt{5}}{2} \\ \frac{\sqrt{5}}{2} \end{bmatrix}$$

$$A^T A \quad 2 \times 2 \quad \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

$$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \quad \text{orthogonal}$$

orthogonal

Apply  $L(\vec{v}_1)$  and  $L(\vec{v}_2)$ . Are they orthogonal?  $A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$   
 $L(\vec{x}) = A\vec{x}$ .

$$L(\vec{v}_1) = \begin{bmatrix} \frac{1}{2} + \frac{\sqrt{5}}{2} \\ 1 \end{bmatrix} \quad L(\vec{v}_2) = \begin{bmatrix} \frac{1}{2} - \frac{\sqrt{5}}{2} \\ 1 \end{bmatrix}$$

orthogonal.

Make  $\vec{v}_1, \vec{v}_2$  unitary, check that  $L(\vec{u}_1)$  is perpendicular to  $L(\vec{u}_2)$ .  
 $\vec{u}_1, \vec{u}_2$

$$\vec{u}_1 = \frac{\sqrt{2}}{5} \begin{bmatrix} \frac{-1+\sqrt{5}}{2} \\ 1 \end{bmatrix} = \begin{bmatrix} \sqrt{1 + \frac{2}{5}} \\ \frac{\sqrt{2}}{\sqrt{5-15}} \end{bmatrix}$$

$$\vec{u}_2 = \frac{\sqrt{2}}{5} \begin{bmatrix} \frac{-1-\sqrt{5}}{2} \\ 1 \end{bmatrix} = \begin{bmatrix} -\sqrt{1 - \frac{2}{5}} \\ \frac{\sqrt{2}}{\sqrt{5+15}} \end{bmatrix}$$

$$L(\vec{u}_1) \cdot L(\vec{u}_2) = (A\vec{u}_1) \cdot (A\vec{u}_2) = (A\vec{u}_1)^T (A\vec{u}_2) = \vec{u}_1^T A^T A \vec{u}_2 =$$

$$= \vec{u}_1^T (\lambda_2 \vec{u}_2) = \lambda_2 \vec{u}_1^T \vec{u}_2 = \lambda_2 \cdot (\vec{u}_1 \cdot \vec{u}_2) = \lambda_2 \cdot 0 = 0.$$

$$\|L(\vec{u}_1)\|^2 = \frac{3}{2} + \frac{\sqrt{5}}{2} = \lambda_1 \quad \leadsto \quad \|L(\vec{u}_1)\| = \sqrt{\lambda_1}$$

$$\|L(\vec{u}_2)\|^2 = \frac{3}{2} - \frac{\sqrt{5}}{2} = \lambda_2 \quad \|L(\vec{u}_2)\| = \sqrt{\lambda_2}$$

The singular values of  $A$  are the square roots of the eigenvalues of  $A^T A$ .

Theorem: Singular value decomposition: Any matrix  $A$  can be decomposed as:  
 $n \times m$

$$A = U \Sigma V^T$$

orthogonal matrix

$$V^{-1} = V^T \quad m \times m$$

$$U^{-1} = U^T \quad n \times n$$

$\Sigma$  is "diagonal", with entries the singular values of  $A$ .  
 $n \times m$

Method:  $A$  is  $n \times m$ ,  $\text{rank}(A) = r$

Find  $\vec{v}_1, \dots, \vec{v}_m$  orthonormal eigenbasis of  $A^T A$ .

$$\begin{array}{cc} \lambda_1 & \lambda_m \\ \sigma_1 & \sigma_m \\ \hline r & m-r \\ \text{non 0} & 0 \end{array}$$

$$\sigma_i = \sqrt{\lambda_i}$$

uppercase  $\Sigma$  sigma  
 lowercase  $\sigma$

$$\vec{u}_1, \dots, \vec{u}_r$$

$$\vec{u}_i = \frac{1}{\sigma_i} A \vec{v}_i$$

$$\vec{u}_{r+1}, \dots, \vec{u}_n$$

$$A = \begin{bmatrix} | & & | \\ \vec{u}_1 & \dots & \vec{u}_r \\ | & & | \end{bmatrix} \begin{bmatrix} \sigma_1 & & 0 \\ & \dots & \\ 0 & & \sigma_r & \dots & 0 \\ & & & \dots & \\ 0 & & & & 0 \end{bmatrix} \begin{bmatrix} -\vec{v}_1- \\ \vdots \\ -\vec{v}_m- \end{bmatrix}$$