

# Lecture 33

## Singular Value Decomposition

Thm: Let  $A$  be an  $n \times m$  matrix, then there exists a factorization

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

where

- $U$  is an  $n \times n$  orthogonal matrix

- $V$  is an  $m \times m$  orthogonal matrix

- $\Sigma$  is "almost diagonal"

$\rightarrow$   $i, j$  of  $\Sigma$  entry is 0 if  $i \neq j$

$\Sigma$  is an  $n \times m$  matrix.

$$\Sigma = \begin{pmatrix} G_1 & & & & \\ & G_2 & & & \\ & & \ddots & & \\ & & & G_n & \\ & & & & 0 \end{pmatrix}$$

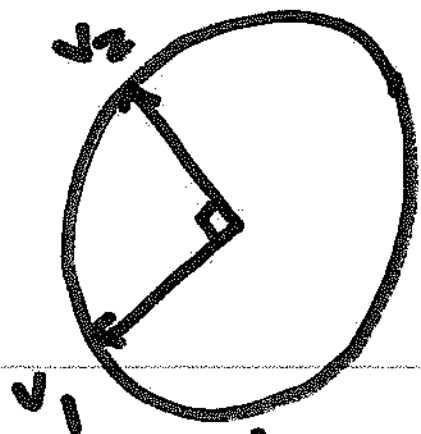
and

the entries  $G_i \geq 0$  and

$$G_1 \geq G_2 \geq G_3 \geq \dots \geq G_n \geq 0.$$

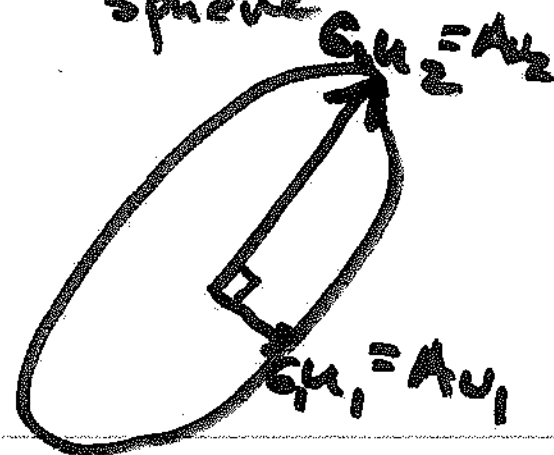
# Conceptualizing the SVD:

Unit sphere  $\{x | x \cdot x = 1\}$



A

Image of unit sphere



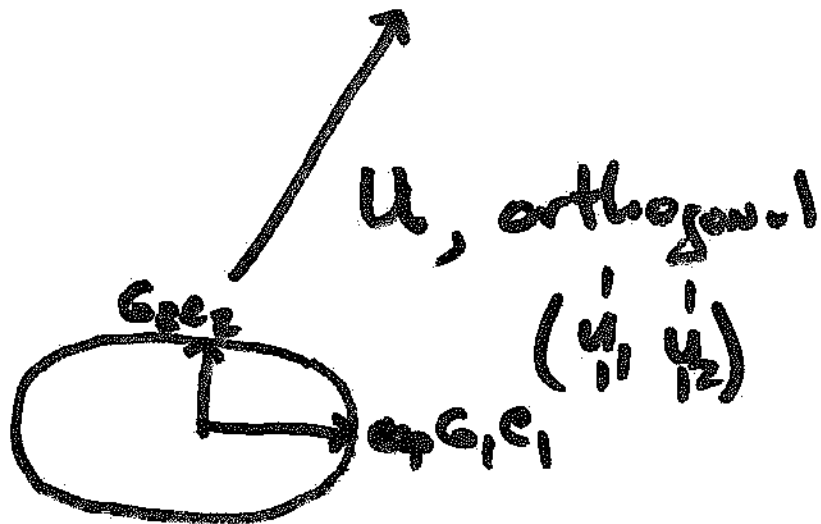
$$V^T = V^{-1}$$

$$\begin{pmatrix} -v_1^T & -v_2^T \\ v_1^T & v_2^T \end{pmatrix}$$



$$\Sigma$$

$$\begin{pmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \end{pmatrix}$$



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

Def: The values

$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n$$

are called the singular values of  $A$ .

The columns of  $U$  are called left singular vectors of  $A$ .

The columns of  $V$  are called right singular vectors of  $A$ .

# How to Find SVD

Want

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

$$\Rightarrow A^T = (U \Sigma V^T)^T = V \Sigma^T U^T$$

$\Sigma$  symmetric

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

$$= V \Sigma^T \cancel{(U^T U)} \Sigma V^T$$

*U is orthogonal*

$$= V (\Sigma^T \Sigma) V^T$$

~~By~~ diagonalizing  $A^T A$  we have:

$$A^T A = V D V^T$$

$$D = \Sigma^T \Sigma$$

$$\Sigma = \begin{bmatrix} \sigma_1 & & & & & \\ & \ddots & & & & \\ & & \sigma_n & & & \\ & & & \ddots & & \\ & & & & & 0 \\ & & & & & & \ddots & & \\ & & & & & & & & 0 \end{bmatrix}$$

$$\Sigma^T = \begin{bmatrix} \sigma_1 & & & & & \\ & \ddots & & & & \\ & & \sigma_n & & & \\ & & & \ddots & & \\ & & & & & 0 \\ & & & & & & \ddots & & \\ & & & & & & & & 0 \end{bmatrix}$$

$$D = \Sigma^T \Sigma = \begin{bmatrix} \sigma_1^2 & & & & & \\ & \ddots & & & & \\ & & \sigma_n^2 & & & \\ & & & \ddots & & \\ & & & & & 0 \\ & & & & & & \ddots & & \\ & & & & & & & & 0 \end{bmatrix}$$

Step 1 for finding SVD:

Diagonalize  $A^T A$

- $V$  is the orthogonal matrix such that  $V D V^T = A^T A$

- $\Sigma = \begin{pmatrix} \sqrt{\lambda_1} & & & & \\ & \ddots & & & \\ & & \sqrt{\lambda_n} & & \\ & & & \ddots & \\ & & & & & 0 \end{pmatrix}$

Square roots of  $n$  largest eigenvalues.

Example:  $A = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix}$ .

Step 1: Compute

$A^T A$  and diagonalize it.

$$A^T A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 2 \end{bmatrix}.$$

① Find eigenvalues

② Find orthonormal basis of  
eigenvectors.



$$P_A(t) = \begin{vmatrix} 1-t & 0 & 1 \\ 0 & 1-t & 1 \\ 1 & 1 & 2-t \end{vmatrix}$$

$$= (1-t) \begin{vmatrix} 1-t & 1 \\ 1 & 2-t \end{vmatrix} +$$

$$\begin{vmatrix} 0 & 1-t \\ 1 & 1 \end{vmatrix}$$

$$= (1-t)^2(2-t) - (1-t) - (1-t)$$

Eigenvalues

$$\lambda = 1, 0, 3.$$

Singular values of  $A$

are

$$\sigma_1 = \sqrt{3}, \quad \sigma_2 = \sqrt{1} = 1$$

! Ignore last  $m-n$  eigenvalues.

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

Find  $V$  st.  $A^T A = V D V^T$

For each  $\lambda$  compute  ~~$\text{Ker}(A - \lambda I)$~~

$\text{Ker}(A^T A - \lambda I)$

Answer:

$$v_1 = \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}$$

$$\lambda_1 = 3, \quad \sigma_1 = \sqrt{3}$$

$$v_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$$

$$\lambda_2 = 1, \quad \sigma_2 = 1$$

$$v_3 = \frac{1}{\sqrt{3}} \begin{bmatrix} -1 \\ -1 \\ 1 \end{bmatrix}$$

$$\lambda_3 = 0$$

$$V = \begin{pmatrix} 1/\sqrt{6} & -1/\sqrt{2} & -1/\sqrt{3} \\ 1/\sqrt{6} & 1/\sqrt{2} & -1/\sqrt{3} \\ 2/\sqrt{6} & 0 & 1/\sqrt{3} \end{pmatrix}.$$

Step 2: Find  $U$ .

For each for first  $n$   
 singular vectors  $v_1, \dots, v_n$   
 compute  $Av_i$

$$u_i = \frac{Av_i}{\|Av_i\|} \quad \Rightarrow \quad \cancel{u_i} = \frac{Av_i}{\sigma_i}$$

! IF  $n > m$ , fill out list by finding  
 basis  $\in \text{IN}(A)^\perp$

$$v_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$u_1 = \frac{Av_1}{\|Av_1\|} = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \frac{1}{\sqrt{2}}$$

$$= \frac{1}{\sqrt{3}\sqrt{2}} \begin{bmatrix} 3 \\ 3 \end{bmatrix} = \frac{1}{\sqrt{18}} \begin{bmatrix} 3 \\ 3 \end{bmatrix}.$$

$$u_2 = \frac{Av_2}{\|Av_2\|} = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} \frac{1}{\sqrt{2}}$$

$$= \begin{bmatrix} -1 \\ 1 \end{bmatrix} \frac{1}{\sqrt{2}}$$

Then  $u$  is

$$u = \begin{bmatrix} 3/\sqrt{10} & -1/\sqrt{2} \\ 3/\sqrt{10} & 1/\sqrt{2} \end{bmatrix}.$$

So

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

Remark:

$$\{ x \cdot x = 1 \}$$



$$A = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

$\text{Sp}(u_1)$



$\text{Sp}(u_2)$