

SVD continued

Recall how the SVD works: A is $m \times n$ of rank r

(1) Compute eigenvalues (w/multiplicity) of $A^T A$:

$$\underbrace{\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_r}_{r = \text{rank}} > \underbrace{\lambda_{r+1} = \dots = \lambda_n = 0}_{n-r = \dim N(A)}$$

(2) Compute orthonormal eigenvectors of $A^T A$

$$\underbrace{\vec{v}_1, \dots, \vec{v}_r}_{\text{basis for Row}(A)}, \underbrace{\vec{v}_{r+1}, \dots, \vec{v}_n}_{\text{basis for } N(A)} \quad A^T A \vec{v}_i = \lambda_i \vec{v}_i$$

(3) Singular values are $\sigma_i = \sqrt{\lambda_i}, \dots, \sigma_r = \sqrt{\lambda_r}$

there are no $\sigma_{r+1}, \dots, \sigma_n$

(4) $\vec{u}_i = \frac{1}{\sigma_i} A \vec{v}_i, \dots, \vec{u}_r = \frac{1}{\sigma_r} A \vec{v}_r$ vs

$\{\vec{u}_1, \dots, \vec{u}_r\}$ is an orthonormal basis for $C(A)$

Vector Form of the SVD: $A = \sigma_1 \vec{u}_1 \vec{v}_1^T + \dots + \sigma_r \vec{u}_r \vec{v}_r^T$

Eg: $A = \begin{bmatrix} 3 & 0 & 3 \\ 4 & 5 & -1 \end{bmatrix}$

(1) $A^T A = \begin{bmatrix} 25 & 20 & 5 \\ 20 & 25 & -5 \\ 5 & -5 & 10 \end{bmatrix} \Rightarrow r=2 \quad \lambda_1=45 \quad \lambda_2=5 \quad \lambda_3=0$

(2) $\lambda=45: A^T A - 45I = \begin{bmatrix} -20 & 20 & 5 \\ 20 & -20 & 5 \\ 5 & -5 & -35 \end{bmatrix} \xrightarrow{\text{ref}} \begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$

$\begin{matrix} x = y \\ y = y \\ z = 0 \end{matrix} \Rightarrow \vec{v}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$

$\lambda=5: A^T A - 5I = \begin{bmatrix} 10 & 20 & 5 \\ 20 & 10 & -5 \\ 5 & -5 & -5 \end{bmatrix} \xrightarrow{\text{ref}} \begin{bmatrix} 1 & 0 & -1/2 \\ 0 & 1 & \sqrt{2} \\ 0 & 0 & 0 \end{bmatrix}$

$x = \frac{1}{2} z$

$\vec{v}_2 = \frac{1}{2} \begin{bmatrix} 1 \\ -4 \\ 2 \end{bmatrix}$

$$\begin{aligned} &\rightarrow y = -1/2z \rightarrow v_2 = \sqrt{6} \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} \\ &\lambda = 0: A^T A \xrightarrow{\text{ref}} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{bmatrix} \rightarrow \begin{matrix} x = -2 \\ y = z \\ z = z \end{matrix} \rightarrow \vec{v}_3 = \frac{1}{\sqrt{3}} \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix} \end{aligned}$$

(3) $\sigma_1 = \sqrt{45}$ $\sigma_2 = \sqrt{5}$ (there is no σ_3)

$$(4) \vec{u}_1 = \frac{1}{\sigma_1} A \vec{v}_1 = \frac{1}{\sqrt{45}} \cdot \frac{1}{\sqrt{2}} \begin{bmatrix} 3 & 0 & 3 \\ 4 & 5 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} = \frac{1}{3\sqrt{10}} \begin{bmatrix} 3 \\ 9 \\ 3 \end{bmatrix} = \frac{1}{\sqrt{10}} \begin{bmatrix} 1 \\ 3 \\ 1 \end{bmatrix}$$

$$\vec{u}_2 = \frac{1}{\sigma_2} A \vec{v}_2 = \frac{1}{\sqrt{15}} \cdot \frac{1}{\sqrt{6}} \begin{bmatrix} 3 & 0 & 3 \\ 4 & 5 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 2 \end{bmatrix} = \frac{1}{3\sqrt{10}} \begin{bmatrix} 9 \\ 9 \\ -3 \end{bmatrix} = \frac{1}{\sqrt{10}} \begin{bmatrix} 3 \\ 3 \\ -1 \end{bmatrix}$$

Note $|\vec{u}_1| = 1$ and $|\vec{u}_2| = 1$

SVD: $A = \sigma_1 \vec{u}_1 \vec{v}_1^T + \sigma_2 \vec{u}_2 \vec{v}_2^T$

$$\begin{bmatrix} 3 & 0 & 3 \\ 4 & 5 & -1 \end{bmatrix} = \sqrt{45} \cdot \frac{1}{\sqrt{10}} \begin{bmatrix} 1 \\ 3 \\ 1 \end{bmatrix} \cdot \frac{1}{\sqrt{2}} [1 \ 1 \ 0] + \sqrt{15} \cdot \frac{1}{\sqrt{10}} \begin{bmatrix} 3 \\ 3 \\ -1 \end{bmatrix} \cdot \frac{1}{\sqrt{6}} [1 \ -1 \ 2]$$

Eg: $A = S$ is symmetric

Let μ_1, \dots, μ_n be the eigenvalues, ordered by $|\mu_1| \geq |\mu_2| \geq \dots \geq |\mu_n|$

Let $\vec{v}_1, \dots, \vec{v}_n$ be an orthonormal eigenbasis, $A \vec{v}_i = \mu_i \vec{v}_i$

(1,2) $A^T A = S^2$ has eigenvalues

$$\lambda_1 = \mu_1^2 \geq \lambda_2 = \mu_2^2 \geq \dots \geq \lambda_n = \mu_n^2$$

with the same eigenvectors.

(3) Singular values are

$$\sigma_1 = \sqrt{\lambda_1} = |\mu_1|, \sigma_2 = \sqrt{\lambda_2} = |\mu_2|, \dots, \sigma_n = \sqrt{\lambda_n} = |\mu_n|$$

$$(4) \vec{u}_i = \frac{1}{\sigma_i} A \vec{v}_i = \frac{1}{|\mu_i|} \mu_i \vec{v}_i = \pm \vec{v}_i \quad (= -\vec{v}_i \text{ if } \mu_i < 0)$$

$$A = \sigma_1 \vec{u}_1 \vec{v}_1^T + \dots + \sigma_n \vec{u}_n \vec{v}_n^T = \mu_1 \vec{v}_1 \vec{v}_1^T + \dots + \mu_n \vec{v}_n \vec{v}_n^T$$

Matrix Form of the SVD

Start with the relations

$$A\vec{v}_i = \sigma_i \vec{u}_i \quad i=1, \dots, r$$

Put into a matrix:

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

We would like these matrices to be square.

Four Fundamental Subspaces

- $\vec{v}_1, \dots, \vec{v}_r$ is an o.n. basis for the row space
 - $\vec{v}_{r+1}, \dots, \vec{v}_n$ is an o.n. basis for $N(A)$
 - $\vec{u}_1, \dots, \vec{u}_r$ is an o.n. basis for $C(A)$
 - **choose** $\vec{u}_{r+1}, \dots, \vec{u}_m$ an o.n. basis for $N(A^T)$
- (compute a basis for $N(A^T)$ by elimination)
(then run Gram-Schmidt)
- $\left. \begin{array}{l} \{\vec{v}_1, \dots, \vec{v}_n\} \\ \text{is an o.n.} \\ \text{basis for } \mathbb{R}^n \end{array} \right\}$
 $\left. \begin{array}{l} \{\vec{u}_1, \dots, \vec{u}_m\} \\ \text{is an o.n.} \\ \text{basis for } \mathbb{R}^m \end{array} \right\}$

Then

$$A \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_r & \vec{v}_{r+1} & \dots & \vec{v}_n \end{bmatrix} = \begin{bmatrix} \sigma_1 \vec{u}_1 & \dots & \sigma_r \vec{u}_r & 0 & \dots & 0 \end{bmatrix}$$

$$= \begin{bmatrix} \vec{u}_1 & \dots & \vec{u}_r & \vec{u}_{r+1} & \dots & \vec{u}_m \end{bmatrix} \begin{bmatrix} \sigma_1 & & & & & \\ & \dots & & & & \\ & & \sigma_r & & & \\ & & & 0 & & \\ & & & & \dots & \\ & & & & & \sigma_r & \\ & & & & & & 0 \end{bmatrix} \leftarrow \begin{array}{l} m \times n \\ \text{matrix} \end{array}$$

Matrix Form of the SVD: $A = U \Sigma V^T$ for

$$U = \begin{bmatrix} \vec{u}_1 & \dots & \vec{u}_m \end{bmatrix} \quad V = \begin{bmatrix} \vec{v}_1 & \dots & \vec{v}_n \end{bmatrix} \quad \Sigma = \begin{bmatrix} \sigma_1 & & & & \\ & \dots & & & \\ & & \sigma_r & & \\ & & & 0 & \\ & & & & \dots & \\ & & & & & \sigma_r & \\ & & & & & & 0 \end{bmatrix}$$

$\leftarrow m \times m$ matrix $\leftarrow n \times n$ matrix $\leftarrow m \times n$ matrix

Procedure for computing the SVD:

(1) Find eigenvalues $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_r > \lambda_{r+1} = \dots = \lambda_n = 0$ of $A^T A$

(2) Find o.n. eigenvectors $\vec{v}_1, \dots, \vec{v}_n$ of $A^T A$

(3) $\sigma_1 = \sqrt{\lambda_1}, \dots, \sigma_r = \sqrt{\lambda_r}$

(4) $\vec{u}_1 = \frac{1}{\sigma_1} A \vec{v}_1, \dots, \vec{u}_r = \frac{1}{\sigma_r} A \vec{v}_r$

(5) Compute a basis of $N(A^T)$ by elimination; run Gram-Schmidt to get an orthonormal basis $\vec{u}_{r+1}, \dots, \vec{u}_m$

$$\hookrightarrow U = \begin{bmatrix} | & & | \\ \vec{u}_1 & \dots & \vec{u}_m \\ | & & | \end{bmatrix} \quad V = \begin{bmatrix} | & & | \\ \vec{v}_1 & \dots & \vec{v}_n \\ | & & | \end{bmatrix} \quad \Sigma = \begin{bmatrix} \sigma_1 & & & & 0 \\ & \dots & & & \\ & & \sigma_r & & \\ 0 & & & \dots & 0 \\ & & & & \sigma_{m-r} & \dots & 0 \\ & & & & & \dots & \\ & & & & & & 0 \dots 0 \end{bmatrix}$$

Eg: $A = \begin{bmatrix} 3 & 0 & 3 \\ 4 & 5 & -1 \end{bmatrix}$ $r=2$ $\vec{v}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$ $\vec{v}_2 = \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$ $\vec{v}_3 = \frac{1}{\sqrt{3}} \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix}$
 $\sigma_1 = \sqrt{45}$ $\sigma_2 = \sqrt{15}$ $\vec{u}_1 = \frac{1}{\sqrt{10}} \begin{bmatrix} 1 \\ 3 \end{bmatrix}$ $\vec{u}_2 = \frac{1}{\sqrt{10}} \begin{bmatrix} 3 \\ -1 \end{bmatrix}$ $N(A^T) = \{0\}$

$$\Rightarrow A = U \Sigma V^T \quad U = \frac{1}{\sqrt{10}} \begin{bmatrix} 1 & 3 \\ 3 & -1 \end{bmatrix} \quad V = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{6} & -1/\sqrt{3} \\ 1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ 0 & 2/\sqrt{6} & 1/\sqrt{3} \end{bmatrix} \quad \Sigma = \begin{bmatrix} \sqrt{45} & 0 & 0 \\ 0 & \sqrt{15} & 0 \end{bmatrix}$$

Eg: $A = S$ is symmetric, as before $\Rightarrow A = U \Sigma V^T$ for

$$U = \begin{bmatrix} | & & | \\ \pm \vec{v}_1 & \dots & \pm \vec{v}_n \\ | & & | \end{bmatrix} \quad V = \begin{bmatrix} | & & | \\ \vec{v}_1 & \dots & \vec{v}_n \\ | & & | \end{bmatrix} \quad \Sigma = \begin{bmatrix} |\mu_1| & & 0 \\ & \dots & \\ 0 & & |\mu_n| \end{bmatrix}$$

In particular, if A is positive (semi) definite then

$$A = U \Sigma V^T \text{ is the same as } A = Q \Lambda Q^T.$$

Symmetry of the SVD

Thm: If $A = U\Sigma V^T$ then $A^T = V\Sigma^T U^T$ is the SVD of A^T .

What is this saying?

- $\vec{u}_1, \dots, \vec{u}_n$ are o.n. eigenvectors of $(A^T)^T A^T = A A^T$ with eigenvalues $\lambda_1, \dots, \lambda_n$. Check:
 $A^T A \vec{v}_i = \lambda_i \vec{v}_i$ and $\vec{u}_i = \frac{1}{\sqrt{\lambda_i}} \vec{v}_i$ ($i=1, \dots, r$)
 $\Rightarrow A A^T \vec{u}_i = \frac{1}{\sqrt{\lambda_i}} A (A^T A \vec{v}_i) = \frac{1}{\sqrt{\lambda_i}} A (\lambda_i \vec{v}_i) = \lambda_i \cdot \frac{1}{\sqrt{\lambda_i}} A \vec{v}_i = \lambda_i \vec{u}_i \checkmark$
For $i > r$, \vec{u}_i is in $N(A^T)$ so $A A^T \vec{u}_i = A \cdot 0 = 0 = \lambda_i \vec{u}_i \checkmark$
- This shows A and A^T have the same singular values
- $A^T \vec{u}_i = \sigma_i \vec{v}_i$ for $i=1, \dots, r$:
 $A^T \vec{u}_i = \frac{1}{\sigma_i} A^T A \vec{v}_i = \frac{\lambda_i}{\sigma_i} \vec{v}_i = \sigma_i \vec{v}_i$ because $\sigma_i^2 = \lambda_i \checkmark$
- $\{\vec{v}_{r+1}, \dots, \vec{v}_n\}$ is a basis for the left nullspace of A^T
= null space of $A \checkmark$

Eg:
$$\begin{bmatrix} 3 & 0 & 3 \\ 4 & 5 & -1 \end{bmatrix} = \frac{1}{\sqrt{10}} \begin{bmatrix} 1 & 3 \\ 3 & -1 \end{bmatrix} \begin{bmatrix} \sqrt{45} & 0 & 0 \\ 0 & \sqrt{15} & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 1/\sqrt{6} & -1/\sqrt{6} & 2/\sqrt{6} \\ -1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{3} \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 3 & 4 \\ 0 & 5 \\ 3 & -1 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{6} & -1/\sqrt{3} \\ 1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ 0 & 2/\sqrt{6} & 1/\sqrt{3} \end{bmatrix} \begin{bmatrix} \sqrt{45} & 0 \\ 0 & \sqrt{15} \\ 0 & 0 \end{bmatrix} \frac{1}{\sqrt{10}} \begin{bmatrix} 1 & 3 \\ 3 & -1 \end{bmatrix}$$