

Linear Algebra Recitation

Week 14

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Singular Value Decomposition $A = W D W^T$ (if A is sym)

True or false: for any matrix A , the matrices $A^T A$, AA^T are symmetric.
 $\in \mathbb{R}^{m \times n}$ $\in \mathbb{R}^{n \times n}$ $\in \mathbb{R}^{m \times m}$

$$(A^T A)^T = A^T (A^T)^T = A^T A$$

$$(AA^T)^T = (A^T)^T A^T = AA^T$$

True or false: for any matrix A , the matrices $A^T A$, AA^T are diagonalizable.

spectral theory of sym. Matrices.

Singular Value Decomposition

True or false: for any matrix A , the eigenvalues of $A^T A$, AA^T are non-negative. $A^T A x = \lambda x$, $\|x\| = 1$

$$\begin{aligned}\underline{\lambda} &= \lambda \|x\|^2 = \lambda (x \cdot x) = (x \cdot \lambda x) = (x \cdot A^T A x) = (Ax \cdot Ax) \\ &= \underline{\|Ax\|^2}\end{aligned}$$

True or false: for any matrix A , the non-negative eigenvalues of $A^T A$, AA^T are the same. *(cf) last week's slides*

$$A^T A x = \lambda x \quad \& \quad \underline{AA^T A x = \lambda Ax}, \quad AA^T y = \lambda y$$

What do you call the square roots of the eigenvalues of $A^T A$ or AA^T ?

Singular Values of A .

Singular Value Decomposition Eval decomp.

The SVD of a matrix $A \in \mathbb{R}^{m \times n}$ is

$$A = W D W^{-1}$$

$$A = U S V^T,$$

where $U \in \mathbb{R}^{m \times m}$ and $V \in \mathbb{R}^{n \times n}$ are **orthogonal** matrices, and $S \in \mathbb{R}^{m \times n}$ contains the singular values on its diagonal. The columns of U are called the “left singular vectors” of A , and the columns of V are called the “right singular vectors” of A .

Question: The eigenvectors of $A^T A$ will become the (left/right) singular vectors of A . The eigenvectors of AA^T will become the (left/right) singular vectors of A .

This is the reason we can find **orthogonal** matrices U, V .

Singular Value Decomposition

How to compute the SVD $A = USV^T$.

- ① Find e.val decomposition of $A^T A$ or AA^T . (Choose the one that is smaller or easier to compute) $\Rightarrow WDW^T$.
- ② Set S to be the “square root” matrix of D . $\Delta S \text{ is not square}$.
- ③ Set W as either left or right singular vector (i.e., $W = U$ or V).
- ④ if $W = V$, find U via the identity $Av_i = \sigma_i u_i$. $AV = US$
- ⑤ if $W = U$, find V via $u_i^T A = \sigma_i v_i^T$. $U^T A = SV^T$

Note 1: once you get U by finding the eigenvectors of AA^T , you **don't** want to compute V by finding eigenvectors of $A^T A$. Instead, use the identity above.

Note 2: if there are singular values that are zero, this means the corresponding singular vectors can be any vector that make U and V orthogonal. Choosing the right vectors in this case require some practice.

Exercises

$$A^T A = A A^T = I \Rightarrow (A^T A - I) = 0$$

Easy.

Find the SVD of $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$.

$$S = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \text{Null space: } \mathbb{R}^2$$

$$A = U S V^T = U I V^T$$

① $U = I, V = I \Rightarrow A = I I I = I$. : Valid

② $U = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, V = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \Rightarrow A = U I V^T = I$: Valid

③ $U = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix}, V = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix}$: Valid

④ $U = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}, V = \begin{bmatrix} 0 & 1 \\ 1 & -1 \end{bmatrix} : A = U I V^T \not\rightarrow$ not an SVD
 $= U^{-1}$ (U, V not orth)
 but is an e.v.al decomp.

Exercises obvious: $A = I \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} I$ is a valid SVD.
 Easy.

Find the SVD of $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$. ① $S = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$

$$\begin{aligned} \text{e.vect of } AA^T &= \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \\ \sigma_1^2 = 1 &\Rightarrow \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \Rightarrow x_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\ \sigma_2^2 = 0 &\Rightarrow \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \Rightarrow x_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \end{aligned}$$

$$U_1^T A = \sigma_1 V_1^T$$

$$\begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} = V_1^T$$

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

$$U_2^T A = \sigma_2 V_2^T$$

$\stackrel{\parallel}{\text{What to do?}}$

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

$$\textcircled{3} V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix},$$

V_2 : chosen to make
 V orthogonal.

$$A = USV^T$$

When A is diagonalizable, the SVD gives the e.val Decomposition.

↳ Singular values $\sigma_i = \text{e.Vals } \lambda_i$

Exercises

Requires some calculation.

$$A^T = \begin{bmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 0 \end{bmatrix} \quad AA^T = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \quad \sigma_1^2 = 3$$

$$\text{Find the SVD of } A = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}. \quad (AA^T - 3I) = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \Rightarrow x_1 = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

$$S = \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \quad (AA^T - I) = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \Rightarrow x_2 = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}$$

$$U = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \quad \underline{U_i^T A} = v_i^T$$

$$V = \begin{bmatrix} \frac{2}{\sqrt{6}} & 0 & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{6}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{3}} \\ \uparrow & \uparrow & \uparrow \\ v_1 & v_2 & v_3 \end{bmatrix} \quad \begin{matrix} \sigma_i \\ (1/\sqrt{2} \ 1/\sqrt{2})/\sqrt{3} \end{matrix} A = \begin{bmatrix} \frac{2}{\sqrt{6}} & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} \end{bmatrix} = v_1^T$$

$$(1/\sqrt{2} \ -1/\sqrt{2}) A = [0 \ -1/\sqrt{2} \ 1/\sqrt{2}] = v_2^T$$

$$A = U S V^T \quad \rightsquigarrow V_3? \text{ Choose any vector that makes } V \text{ orthogonal: } V_3^T = \left[\frac{1}{\sqrt{3}}, -\frac{1}{\sqrt{3}}, -\frac{1}{\sqrt{3}} \right]$$

Exercises

$$A = \begin{bmatrix} 2 & -1 \\ 2 & 1 \end{bmatrix} \quad A^T A = \begin{bmatrix} 5 & 3 \\ 3 & 5 \end{bmatrix}$$

Requires some calculation.

Find the SVD of $A = \begin{bmatrix} 2 & 2 \\ -1 & 1 \end{bmatrix}$. $A A^T = \begin{bmatrix} 8 & 0 \\ 0 & 2 \end{bmatrix}$ ← choose this one!

$$\sigma_1^2 = 8 \Rightarrow A A^T - 8I = \begin{bmatrix} 0 & 0 \\ 0 & -6 \end{bmatrix} \Rightarrow x_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \text{①} \Rightarrow U = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, S = \begin{bmatrix} 2\sqrt{2} & 0 \\ 0 & \sqrt{2} \end{bmatrix}$$

$$\sigma_2^2 = 2 \Rightarrow A A^T - 2I = \begin{bmatrix} 6 & 0 \\ 0 & 0 \end{bmatrix} \Rightarrow x_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad \text{②}$$

$$\text{Use } U_i^T A = \sigma_i v_i^T$$

$$\begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} 2 & 2 \\ -1 & 1 \end{bmatrix} \frac{1}{2\sqrt{2}} = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} = v_1^T \quad \text{③} \Rightarrow V = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}$$

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

$$\underline{\underline{A = U S V^T}}$$

Exercises

Conceptual.

If A is an orthogonal matrix, what are the singular values of A ?

$$A^T A = I$$

$$\Rightarrow \sigma_i^2 = 1 \text{ for all } i$$

We choose the positive ones $\sigma_i = 1^{\top} v_i$ (Convention)

Exercises

Conceptual.

If $A = USV^T$, what is the SVD of A^T ?

$$A^T = (USV^T)^T = VS^T U^T = VSU^T$$

Furthermore, if A is invertible, what is the SVD of A^{-1} ?

$$A^{-1} = (VSU^T)^{-1} = V S^{-1} U^T$$

$$S^{-1} = \begin{bmatrix} \frac{1}{\sigma_1} & & \\ & \ddots & \\ & & \frac{1}{\sigma_n} \end{bmatrix}$$