

# Lecture 4 The Singular Value Decomposition

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## 1 Prerequisite

**Lemma 1.1.** *Given symmetric matrix  $\mathbf{A}$ , then the eigenvalues of  $\mathbf{A}$  are real.*

**Theorem 1.1.** *Given symmetric matrix  $\mathbf{A}$ , then  $\mathbf{A}$  can be factored as*

$$\mathbf{A} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^*,$$

where

- $\mathbf{Q}$  is unitary;
- $\mathbf{\Lambda}$  is diagonal, with the eigenvalues of  $\mathbf{A}$  on its diagonal.

*Proof.* By induction of the dimension of  $\mathbf{A}$ . □

## 2 Solutions

### 2.1 Exercise 4.1(e)

First we compute the singular values  $\sigma_i$  by finding the eigenvalues of  $\mathbf{A}^*\mathbf{A}$ :

$$\mathbf{A}^*\mathbf{A} = \begin{pmatrix} 2 & 2 \\ 2 & 2 \end{pmatrix},$$

the characteristic polynomial of  $\mathbf{A}^*\mathbf{A}$  is

$$\det(\mathbf{A}^*\mathbf{A} - \lambda\mathbf{I}) = \lambda(\lambda - 4) = 0,$$

so the singular values are  $\sigma_1 = 0, \sigma_2 = 2$ . For  $\lambda = 4$ , we have

$$\mathbf{A}^*\mathbf{A} - 4\mathbf{I} = \begin{pmatrix} -2 & 2 \\ 2 & -2 \end{pmatrix},$$

a unit vector in the kernel of the matrix is  $\mathbf{v}_2 = (1/\sqrt{2}, 1/\sqrt{2})^T$ . For  $\lambda = 0$ , we have

$$\mathbf{A}^*\mathbf{A} - 0\mathbf{I} = \begin{pmatrix} 2 & 2 \\ 2 & 2 \end{pmatrix},$$

a unit vector in the kernel of the matrix is  $\mathbf{v}_2 = (-1/\sqrt{2}, 1/\sqrt{2})^T$ . So at this point we know that

$$\mathbf{A} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^* = (\mathbf{u}_1, \mathbf{u}_2) \begin{pmatrix} 2 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix}.$$

Finally we can compute  $\mathbf{u}_1$  by the formula  $\sigma_i \mathbf{u}_i = \mathbf{A}\mathbf{v}_i$ , this gives  $\mathbf{u}_i = (\sqrt{2}/2, \sqrt{2}/2)$ , then by  $\mathbf{u}_2^* \mathbf{u}_1 = 0$  and  $\|\mathbf{u}_2\|_2 = 1$  we can get a  $\mathbf{u}_2 = (-\sqrt{2}/2, \sqrt{2}/2)$ . So in this full glory the SVD is

$$\mathbf{A} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^* = \begin{pmatrix} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^*.$$

## 2.2 Exercise 4.2

Assume that

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} = \begin{pmatrix} \alpha_1^T \\ \alpha_1^T \\ \vdots \\ \alpha_m^T \end{pmatrix}$$

then we can get that matrix  $\mathbf{B}$

$$\mathbf{B} = \begin{pmatrix} a_{m1} & \cdots & a_{21} & a_{1n} \\ a_{m2} & \cdots & a_{22} & a_{12} \\ \vdots & \vdots & \ddots & \vdots \\ a_{mn} & \cdots & a_{2n} & a_{1n} \end{pmatrix} = (\alpha_m, \cdots, \alpha_2, \alpha_1)$$

that is

$$\mathbf{B} = \mathbf{A}^T \begin{pmatrix} 0 & 0 & \cdots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 1 & \cdots & 0 \\ 1 & 0 & \cdots & 0 \end{pmatrix} = \mathbf{A}^T \mathbf{P}$$

it is clear that  $\mathbf{P}$  is a orthogonal matrix since  $\mathbf{P}\mathbf{P}^T = \mathbf{P}^T\mathbf{P} = \mathbf{I}_m$ , then

$$\mathbf{B}\mathbf{B}^T = \mathbf{A}^T \mathbf{P}\mathbf{P}^T \mathbf{A} = \mathbf{A}^T \mathbf{A},$$

which means that  $\mathbf{B}$  and  $\mathbf{A}$  have that same singular values.

## 2.3 Exercise 4.3

See [matlab code](#).

## 2.4 Exercise 4.4

If  $\mathbf{A}, \mathbf{B}$  are unitary equivalent, we can get that  $\mathbf{A}, \mathbf{B}$  have the same singular values by the same argument in 4.2. It is clear that  $\mathbf{I}, -\mathbf{I}$  have the same singular values, but  $\mathbf{I}$  and  $-\mathbf{I}$  can't be unitary equivalent since  $\mathbf{I} \neq \mathbf{Q}(-\mathbf{I})\mathbf{Q}^* = -\mathbf{I}$ .

## 2.5 Exercise 4.5

If  $\mathbf{A}$  is a real matrix, then by [Theorem 1.1](#),  $\mathbf{A}^*\mathbf{A}$  is a real symmetric matrix and  $\mathbf{A}^*\mathbf{A}$  has a real eigen decomposition

$$\mathbf{A}^*\mathbf{A} = \mathbf{V}\mathbf{\Lambda}\mathbf{V}^*$$

where  $\mathbf{\Lambda}$  is a diagonal matrix with its entries are eigenvalues of  $\mathbf{A}^*\mathbf{A}$ , then we can get  $\mathbf{U}$  by  $\mathbf{A}\mathbf{v}_i = \sigma_i\mathbf{u}_i$ , we get the real SVD of  $\mathbf{A}$ :

$$\mathbf{A} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^*.$$