Lecture 4 The Singular Value Decomposition

Fang Zhu

April 20, 2023

1 Prerequisite

Lemma 1.1. Given summetrix matrix A, then the eigenvalues of A are real.

Theorem 1.1. Given symmetric matrix A, then A can be factored as

$$A = Q\Lambda Q^{\star}$$
,

where

- Q is unitary;
- ullet Λ is diagonal, with the eigenvalues of $oldsymbol{A}$ on its dagonal.

Proof. By induction of the dimension of A.

2 Solutions

2.1 Exercise 4.1(e)

First we compute the singular values σ_i by finding the eigenvalues of A^*A :

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

the characteristic polynomial of A^*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

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

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

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

a unit vector in the kernel of the matrix is $v_2 = (-1/\sqrt{2}, 1/\sqrt{2})^T$. So at this point we know that

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

Finally we can compute u_1 by the formula $\sigma_i u_i = A v_i$, this gives $u_i = (\sqrt{2}/2, \sqrt{2}/2)$, then by $u_2^* u_1 = 0$ and $||u_2||_2 = 1$ we can get a $u_2 = (-\sqrt{2}/2, \sqrt{2}/2)$. So in this full glory the SVD is

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

2.2 Exercise 4.2

Assume that

$$oldsymbol{A} = egin{pmatrix} a_{11} & a_{12} & \cdots a_{1n} \ a_{21} & a_{22} & \cdots a_{2n} \ dots & dots & \ddots dots \ a_{m1} & a_{m2} & \cdots a_{mn} \end{pmatrix} = egin{pmatrix} oldsymbol{lpha}_1^T \ oldsymbol{lpha}_1^T \ dots \ oldsymbol{lpha}_m^T \end{pmatrix}$$

then we can get that matrix \boldsymbol{B}

$$\boldsymbol{B} = \begin{pmatrix} a_{m1} & \cdots & a_{21} & a_{1n} \\ a_{m2} & \cdots & a_{22} & a_{12} \\ \vdots & \vdots & \ddots \vdots & \\ a_{mn} & \cdots & a_{2n} & a_{1n} \end{pmatrix} = (\boldsymbol{\alpha}_m, \cdots, \boldsymbol{\alpha}_2, \boldsymbol{\alpha}_1)$$

that is

$$m{B} = m{A}^T egin{pmatrix} 0 & 0 & \cdots 1 \ dots & dots & \ddots dots \ 0 & 1 & \cdots 0 \ 1 & 0 & \cdots 0 \end{pmatrix} = m{A}^T m{P}$$

it is clear that P is a othogonomal matrix since $PP^T = P^TP = I_m$, then

$$BB^T = A^T P P^T A = A^T A.$$

which means that B and A have that same signlar values.

2.3 Exercise 4.3

```
[U,S,V] = svd(A);
6
   [m, n] = size(A); % U:m*m, V:n*n
8
9
  ss = diag(S);
  v_end_points_xsub = V(1, :);
11
12
  v_end_points_ysub = V(2, :);
13
14 \mid u_{end_points_xsub} = U(1, :);
15 | u_end_points_ysub = U(2, :);
16
17 | % Plot a circle with radius = 1
18 | ths = linspace(0, 2*pi, 100);
19 \mid x = \cos(ths);
20 \mid y = \sin(ths);
21
  plot(x, y);
  axis equal;
23 hold on;
24
25 | % Plot the ellipse transformed by A
26 | xy = [x;y];
  transformed_xy = A*xy;
   transformed_x = transformed_xy(1,:);
  transformed_y = transformed_xy(2,:);
  plot(transformed_x, transformed_y);
31
  hold on;
32
34
35 | % Plot the right singular vector v1, v2
   quiver(0,0,v_end_points_xsub(1),v_end_points_ysub(1),
       "AutoScale", "off");
37
  hold on;
   quiver(0,0,v_end_points_xsub(2),v_end_points_ysub(2),
       "AutoScale", "off");
39
40 |% Plot the left singular vector u1, u2 scaled by the
       singular vales
41
   quiver (0,0,ss(1)*u_end_points_xsub(1),ss(1)*
      u_end_points_ysub(1), "AutoScale", "off");
42
  hold on;
   quiver(0,0,ss(2)*u_end_points_xsub(2),ss(2)*
      u_end_points_ysub(2), "AutoScale", "off");
```

2.4 Exercise 4.4

If A, B are unitary quivalent, we can use the same argument as in Exercise 4.2 to show that they have the same singular values. It is evident that the matrices I and -I have the same singular values. However, I and -I can not be unitary equivalent because I is not equal to $Q(-I)Q^*$, where Q is unitary matrix, and thus, the latter is equal to -I.

2.5 Exercise 4.5

By Theorem 1.1, if A is a real matrix, then A^*A is a real symmetric matrix with a real eigen decomposition:

$$A^*A = V\Lambda V^*$$

where Λ is a diagonal matrix whose entries are the eigenvalues of A^*A . We can obtain U by solving $Av_i = \sigma_i u_i$, which gives us the real singular value decomposition(SVD) of A:

$$A = U\Sigma V^*$$
,

where Σ is a diagonal matrix whose entries are the singular values of A.