

Homework Set 9 (Lecture 28)

Problem 1

Example 14. Compute the SVD of $A = \begin{bmatrix} -3 & 4 \\ -5 & 0 \end{bmatrix}$. That is, decompose A as $A = U\Sigma V^T$.

Solution. (by hand; you will need to show all steps on the final exam)

- First, we need to diagonalize $A^T A = \begin{bmatrix} -3 & -5 \\ 4 & 0 \end{bmatrix} \begin{bmatrix} -3 & 4 \\ -5 & 0 \end{bmatrix} = \begin{bmatrix} 34 & -12 \\ -12 & 16 \end{bmatrix}$.
 $\det\left(\begin{bmatrix} 34-\lambda & -12 \\ -12 & 16-\lambda \end{bmatrix}\right) = (34-\lambda)(16-\lambda) - 144 = \lambda^2 - 50\lambda + 400 = (\lambda-10)(\lambda-40)$
Hence, the eigenvalues of $A^T A$ are 10, 40.

- $\lambda = 10$: $\begin{bmatrix} 24 & -12 \\ -12 & 6 \end{bmatrix} \xrightarrow{R_2 + \frac{1}{2}R_1 \Rightarrow R_2} \begin{bmatrix} 24 & -12 \\ 0 & 0 \end{bmatrix} \xrightarrow{\frac{1}{24}R_1 \Rightarrow R_1} \begin{bmatrix} 1 & -\frac{1}{2} \\ 0 & 0 \end{bmatrix}$
Hence, the 10-eigenspace has basis $\begin{bmatrix} 1/2 \\ 1 \end{bmatrix}$ or, easier for working by hand, $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$.

- $\lambda = 40$: $\begin{bmatrix} -6 & -12 \\ -12 & -24 \end{bmatrix} \xrightarrow{R_2 - 2R_1 \Rightarrow R_2} \begin{bmatrix} -6 & -12 \\ 0 & 0 \end{bmatrix} \xrightarrow{-\frac{1}{6}R_1 \Rightarrow R_1} \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix}$
Hence, the 40-eigenspace has basis $\begin{bmatrix} -2 \\ 1 \end{bmatrix}$.

Thus $A^T A = P D P^T$ with $D = \begin{bmatrix} 40 & \\ & 10 \end{bmatrix}$ and $P = \frac{1}{\sqrt{5}} \begin{bmatrix} -2 & 1 \\ 1 & 2 \end{bmatrix}$.

[We have to normalize the eigenvectors! Otherwise, we would only have a diagonalization $P D P^{-1}$.]

- Since $A^T A = V \Sigma^2 V^T$, we conclude that $V = \frac{1}{\sqrt{5}} \begin{bmatrix} -2 & 1 \\ 1 & 2 \end{bmatrix}$ and $\Sigma = \begin{bmatrix} \sqrt{40} & \\ & \sqrt{10} \end{bmatrix}$.
- From $A v_i = \sigma_i u_i$, we find $u_1 = \frac{1}{\sigma_1} A v_1 = \frac{1}{\sqrt{40}} \begin{bmatrix} -3 & 4 \\ -5 & 0 \end{bmatrix} \frac{1}{\sqrt{5}} \begin{bmatrix} -2 \\ 1 \end{bmatrix} = \frac{1}{\sqrt{200}} \begin{bmatrix} 10 \\ 10 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$.
Likewise, $u_2 = \frac{1}{\sigma_2} A v_2 = \frac{1}{\sqrt{10}} \begin{bmatrix} -3 & 4 \\ -5 & 0 \end{bmatrix} \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \frac{1}{\sqrt{50}} \begin{bmatrix} 5 \\ -5 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$. Hence, $U = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$.

In summary, $A = U \Sigma V^T$ with $U = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$, $\Sigma = \begin{bmatrix} \sqrt{40} & \\ & \sqrt{10} \end{bmatrix}$, $V = \frac{1}{\sqrt{5}} \begin{bmatrix} -2 & 1 \\ 1 & 2 \end{bmatrix}$.

Solution. (using Sage) We obtain the same solution (up to a sign in U and V):

```
>>> A = matrix(RDF, [[-3,4],[-5,0]])
>>> U,S,V = A.SVD()
>>> U
(
  -0.7071067811865475  -0.7071067811865476
  -0.7071067811865476   0.7071067811865475
)
>>> S
(
  6.324555320336758          0.0
          0.0  3.16227766016838
)
>>> V
(
  0.8944271909999159  -0.44721359549995804
  -0.44721359549995804  -0.8944271909999159
)
```