

## Homework 11

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1. Find the singular value decomposition of the following two matrices.

(a)

$$\begin{bmatrix} 1 & -2 \\ -3 & 6 \end{bmatrix}$$

$$\text{Let } A = \begin{bmatrix} 1 & -2 \\ -3 & 6 \end{bmatrix}.$$

$$\begin{aligned} A^T A &= \begin{bmatrix} 1 & -3 \\ -2 & 6 \end{bmatrix} \begin{bmatrix} 1 & -2 \\ -3 & 6 \end{bmatrix} = \begin{bmatrix} 10 & -20 \\ -20 & 40 \end{bmatrix} \\ \det(A^T A - \lambda I) &= \begin{vmatrix} 10 - \lambda & -20 \\ -20 & 40 - \lambda \end{vmatrix} = 0 \\ &= (10 - \lambda)(40 - \lambda) - (-20)(-20) = 0 \\ &= \lambda^2 - 50\lambda = 0 \\ &= \lambda(\lambda - 50) = 0 \\ &\therefore \lambda = 0, 50 \end{aligned}$$

$\lambda_1 = 0$  and  $\lambda_2 = 50$  are the eigenvalues of  $A^T A$ .

$$\begin{aligned} \lambda_1 = 0 &\implies \sigma_1 = 0 \\ \lambda_2 = 50 &\implies \sigma_2 = \sqrt{50} = 5\sqrt{2} \end{aligned}$$

$$\therefore \Sigma = \begin{bmatrix} 0 & 0 \\ 0 & 5\sqrt{2} \end{bmatrix}$$

And so:

$$\begin{aligned} \text{rref}(A^T A - \lambda_1 I) &= \text{rref} \begin{bmatrix} 10 - 0 & -20 \\ -20 & 40 - 0 \end{bmatrix} = \begin{bmatrix} 1 & -2 \\ 0 & 0 \end{bmatrix} \\ \therefore \ker(A^T A - \lambda_1 I) &= \ker \begin{bmatrix} 1 & -2 \\ 0 & 0 \end{bmatrix} = \text{span} \left\{ \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right\} \end{aligned}$$

As such,  $\vec{v}_1 = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 1 \end{bmatrix}$  is an eigenvector corresponding to  $\lambda_1 = 0$ .

$$\begin{aligned} \text{rref}(A^T A - \lambda_2 I) &= \text{rref} \begin{bmatrix} 10 - 50 & -20 \\ -20 & 40 - 50 \end{bmatrix} = \begin{bmatrix} 1 & 1/2 \\ 0 & 0 \end{bmatrix} \\ \therefore \ker(A^T A - \lambda_2 I) &= \ker \begin{bmatrix} 1 & 1/2 \\ 0 & 0 \end{bmatrix} = \text{span} \left\{ \begin{bmatrix} -1/2 \\ 1 \end{bmatrix} \right\} \end{aligned}$$

As such,  $\vec{v}_2 = \frac{2}{\sqrt{5}} \begin{bmatrix} -1/2 \\ 1 \end{bmatrix}$  is an eigenvector corresponding to  $\lambda_2 = 50$ .

$$\therefore V = \begin{bmatrix} 2/\sqrt{5} & -1/\sqrt{5} \\ 1/\sqrt{5} & 2/\sqrt{5} \end{bmatrix} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & -1 \\ 1 & 2 \end{bmatrix}$$

Then,  $\vec{\mathbf{u}}_i = \frac{1}{\sigma_i} A \vec{\mathbf{v}}_i$  for  $i = 1, 2$ :

Note  $\sigma_1 = 0$ , so we move on to  $\vec{\mathbf{u}}_2$ .

$$\begin{aligned}\vec{\mathbf{u}}_2 &= \frac{1}{5\sqrt{2}} \begin{bmatrix} 1 & -2 \\ -3 & 6 \end{bmatrix} \frac{2}{\sqrt{5}} \begin{bmatrix} -1/2 \\ 1 \end{bmatrix} \\ &= \frac{1}{\sqrt{10}} \begin{bmatrix} -1 \\ 3 \end{bmatrix}\end{aligned}$$

Since  $\sigma_1 = 0$  and we need one more vector, we need to choose a vector  $\vec{\mathbf{u}}_1$  such that it is orthogonal to  $\vec{\mathbf{u}}_2$ . Since it's just  $\mathbb{R}^2$ , we can just eyeball and make  $\vec{\mathbf{u}}_1 = \frac{1}{\sqrt{10}} \begin{bmatrix} 3 \\ 1 \end{bmatrix}$ .

Sanity check:  $\langle \vec{\mathbf{u}}_1, \vec{\mathbf{u}}_2 \rangle = \frac{1}{\sqrt{10}}((-1)(3) + 3(1)) = 0$ .

$$\therefore U = \begin{bmatrix} 3/\sqrt{10} & -1/\sqrt{10} \\ 1/\sqrt{10} & 3/\sqrt{10} \end{bmatrix} = \frac{1}{\sqrt{10}} \begin{bmatrix} 3 & -1 \\ 1 & 3 \end{bmatrix}$$

And so, we have that the SVD for  $A = \begin{bmatrix} 1 & -2 \\ -3 & 6 \end{bmatrix}$  is:

$$\begin{aligned}A &= U \Sigma V^\top \\ &= \begin{bmatrix} 3/\sqrt{10} & -1/\sqrt{10} \\ 1/\sqrt{10} & 3/\sqrt{10} \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 5\sqrt{2} \end{bmatrix} \begin{bmatrix} 2/\sqrt{5} & -1/\sqrt{5} \\ 1/\sqrt{5} & 2/\sqrt{5} \end{bmatrix}^\top \\ &= \frac{1}{\sqrt{10}} \begin{bmatrix} 3 & -1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 5\sqrt{2} \end{bmatrix} \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & 1 \\ -1 & 2 \end{bmatrix} \\ &= \begin{bmatrix} 1 & -2 \\ -3 & 6 \end{bmatrix}\end{aligned}$$

(b)

$$\begin{bmatrix} 2 & 1 & 0 & -1 \\ 0 & -1 & 1 & -1 \end{bmatrix}$$

$$\text{Let } A = \begin{bmatrix} 2 & 1 & 0 & -1 \\ 0 & -1 & 1 & -1 \end{bmatrix}.$$

$$\begin{aligned}A^\top A &= \begin{bmatrix} 4 & 2 & 0 & -2 \\ 2 & 2 & -1 & 0 \\ 0 & -1 & 1 & -1 \\ -2 & 0 & -1 & 2 \end{bmatrix} \\ \det(A^\top A - \lambda I) &= \begin{vmatrix} 4-\lambda & 2 & 0 & -2 \\ 2 & 2-\lambda & -1 & 0 \\ 0 & -1 & 1-\lambda & -1 \\ -2 & 0 & -1 & 2-\lambda \end{vmatrix} = 0 \\ &= \lambda^4 - 9\lambda^3 + 18\lambda^2 = 0 \\ &= \lambda^2(\lambda - 3)(\lambda - 6) = 0 \\ \therefore \lambda &= 0, 3, 6\end{aligned}$$

$\lambda_1 = 0$ ,  $\lambda_2 = 3$ , and  $\lambda_3 = 6$  are eigenvalues of  $A^\top A$ .

$$\begin{aligned}\lambda_1 = 0 &\implies \sigma_1 = 0 \\ \lambda_2 = 3 &\implies \sigma_2 = \sqrt{3} \\ \lambda_3 = 6 &\implies \sigma_3 = \sqrt{6}\end{aligned}$$

$$\therefore \Sigma = \begin{bmatrix} \sqrt{3} & 0 & 0 & 0 \\ 0 & \sqrt{6} & 0 & 0 \end{bmatrix}$$

And so:

$$\begin{aligned}\text{rref}(A^\top A - \lambda_1 I) &= \text{rref} \begin{bmatrix} 4-0 & 2 & 0 & -2 \\ 2 & 2-0 & -1 & 0 \\ 0 & -1 & 1-0 & -1 \\ -2 & 0 & -1 & 2-0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1/2 & -1 \\ 0 & 1 & -1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \\ \therefore \ker(A^\top A - \lambda_1 I) &= \ker \begin{bmatrix} 1 & 0 & 1/2 & -1 \\ 0 & 1 & -1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \text{span} \left\{ \begin{bmatrix} -1/2 \\ 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 0 \\ 1 \end{bmatrix} \right\}\end{aligned}$$

As such,  $\frac{2}{3} \begin{bmatrix} -1/2 \\ 1 \\ 1 \\ 0 \end{bmatrix}$ ,  $\frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ -1 \\ 0 \\ 1 \end{bmatrix}$  are eigenvectors corresponding to  $\lambda_1 = 0$ .

$$\begin{aligned}\text{rref}(A^\top A - \lambda_2 I) &= \text{rref} \begin{bmatrix} 4-3 & 2 & 0 & -2 \\ 2 & 2-3 & -1 & 0 \\ 0 & -1 & 1-3 & -1 \\ -2 & 0 & -1 & 2-3 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \\ \therefore \ker(A^\top A - \lambda_2 I) &= \ker \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \text{span} \left\{ \begin{bmatrix} 0 \\ 1 \\ -1 \\ 1 \end{bmatrix} \right\}\end{aligned}$$

As such,  $\vec{v}_2 = \frac{1}{\sqrt{3}} \begin{bmatrix} 0 \\ 1 \\ -1 \\ 1 \end{bmatrix}$  is an eigenvector corresponding to  $\lambda_2 = 3$ .

$$\begin{aligned}\text{rref}(A^\top A - \lambda_3 I) &= \text{rref} \begin{bmatrix} 4-6 & 2 & 0 & -2 \\ 2 & 2-6 & -1 & 0 \\ 0 & -1 & 1-6 & -1 \\ -2 & 0 & -1 & 2-6 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \\ \therefore \ker(A^\top A - \lambda_3 I) &= \ker \begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \text{span} \left\{ \begin{bmatrix} -2 \\ -1 \\ 0 \\ 1 \end{bmatrix} \right\}\end{aligned}$$

As such,  $\vec{v}_3 = \frac{1}{\sqrt{6}} \begin{bmatrix} -2 \\ -1 \\ 0 \\ 1 \end{bmatrix}$  is an eigenvector corresponding to  $\lambda_3 = 6$ .

$$\therefore V = \begin{bmatrix} -1/3 & 1/\sqrt{3} & 0 & -\sqrt{2/3} \\ 2/3 & -1/\sqrt{3} & 1/\sqrt{3} & -1/\sqrt{6} \\ 2/3 & 0 & -1/\sqrt{3} & 0 \\ 0 & 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{6} \end{bmatrix}$$

Then,  $\vec{u}_i = \frac{1}{\sigma_i} A \vec{v}_i$  for  $i = 2, 3$ :

$$\vec{u}_2 = \frac{1}{\sqrt{3}} \begin{bmatrix} 2 & 1 & 0 & -1 \\ 0 & -1 & 1 & -1 \end{bmatrix} \frac{1}{\sqrt{3}} \begin{bmatrix} 0 \\ 1 \\ -1 \\ 1 \end{bmatrix}$$

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

$$\vec{u}_3 = \frac{1}{\sqrt{6}} \begin{bmatrix} 2 & 1 & 0 & -1 \\ 0 & -1 & 1 & -1 \end{bmatrix} \frac{1}{\sqrt{6}} \begin{bmatrix} -2 \\ -1 \\ 0 \\ 1 \end{bmatrix}$$

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

$$\therefore U = \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$$

And so, we have that the SVD for  $A = \begin{bmatrix} 1 & -2 \\ -3 & 6 \end{bmatrix}$  is:

$$A = U \Sigma V^T$$

$$= \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} \sqrt{3} & 0 & 0 & 0 \\ 0 & \sqrt{6} & 0 & 0 \end{bmatrix} \begin{bmatrix} -1/3 & 1/\sqrt{3} & 0 & -\sqrt{2/3} \\ 2/3 & -1/\sqrt{3} & 1/\sqrt{3} & -1/\sqrt{6} \\ 2/3 & 0 & -1/\sqrt{3} & 0 \\ 0 & 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{6} \end{bmatrix}^T$$

$$= \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} \sqrt{3} & 0 & 0 & 0 \\ 0 & \sqrt{6} & 0 & 0 \end{bmatrix} \begin{bmatrix} -1/3 & 2/3 & 2/3 & 0 \\ 1/\sqrt{3} & -1/\sqrt{3} & 0 & 1/\sqrt{3} \\ 0 & 1/\sqrt{3} & -1/\sqrt{3} & 1/\sqrt{3} \\ -\sqrt{2/3} & -1/\sqrt{6} & 0 & 1/\sqrt{6} \end{bmatrix}$$

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

**2. A famous application of spectral theorem and SVD is the spectral graph theory. A graph  $(V, E)$  is a set of vertices  $V$  with edge set  $E$ . We say that for  $x, y \in V$ ,  $x \sim y$  if  $x$  and  $y$  are connected by an edge in  $E$ . The Graph Laplacian is defined to be a matrix  $A$  of size  $|V| \times |V|$**

$$A_{x,y} = \begin{cases} 1 & \text{if } x \sim y \\ -(\text{number of edges starting from } x) & \text{if } x = y \\ 0 & \text{otherwise} \end{cases}$$

For example, a triangle with three vertices

$$A = \begin{bmatrix} -2 & 1 & 1 \\ 1 & -2 & 1 \\ 1 & 1 & -2 \end{bmatrix}$$

**(a) Find the SVD for the Laplacian of the triangle graph.**

Using an online calculator, we find  $A = U\Sigma V^T$  for

$$U = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{6} & 1/\sqrt{3} \\ 0 & -\sqrt{2/3} & 1/\sqrt{3} \\ -1/\sqrt{2} & 1/\sqrt{6} & 1/\sqrt{3} \end{bmatrix}, \quad \Sigma = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad V = \begin{bmatrix} -1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ 0 & \sqrt{2/3} & 1/\sqrt{3} \\ 1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \end{bmatrix}.$$

**(b) Write down the graph Laplacian matrix for the square graph and then find its SVD.**

Using the definition above, the Laplacian matrix  $A$  for a square graph is:

$$A = \begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ -1 & 0 & -1 & 2 \end{bmatrix}$$

Again, sparing the computation. The SVD for the square graph  $A = U\Sigma V^T$  is given by:

$$U = V = \begin{bmatrix} -1/2 & 0 & -1/\sqrt{2} & 1/2 \\ 1/2 & -1/\sqrt{2} & 0 & 1/2 \\ -1/2 & 0 & 1/\sqrt{2} & 1/2 \\ 1/2 & 1/\sqrt{2} & 0 & 1/2 \end{bmatrix}, \quad \Sigma = \begin{bmatrix} 4 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$