

## MAT 167: Advanced Linear Algebra Solutions to SVD Exercises in HW8

**Exercise 4.1** I already described the strategy for computing SVD of these small matrices via eigenvalue/eigenvector computation of  $A^T A$ .

(a)  $\begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix}$ . First compute  $A^T A$ , which is  $\begin{bmatrix} 9 & 0 \\ 0 & 4 \end{bmatrix}$ . The characteristic equation is:

$$\det(A^T A - \lambda I) = \det\left(\begin{bmatrix} 9 - \lambda & 0 \\ 0 & 4 - \lambda \end{bmatrix}\right) = (9 - \lambda)(4 - \lambda) = 0.$$

Thus, the eigenvalues are:  $\lambda = 9, 4$ . Therefore, we immediately get the singular values of  $A$  as  $\sigma_1 = 3$ , and  $\sigma_2 = 2$ . Now, we need to compute the eigenvectors. To do so, need to solve:

$$A^T A \mathbf{v}_j = \lambda_j \mathbf{v}_j, \quad \text{or} \quad (A^T A - \lambda_j I) \mathbf{v}_j = \mathbf{0}, \quad j = 1, 2.$$

Let  $\mathbf{v}_1 = \begin{bmatrix} x & y \end{bmatrix}^T$ .

$$\begin{bmatrix} 9 - 9 & 0 \\ 0 & 4 - 9 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ -5y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

Therefore  $y = 0$  is a must.  $x$  can be chosen arbitrary if this is just for eigenvector computation, but we have additional constraint. We are computing the column vectors of the unitary (or orthogonal) matrix  $V$  of the SVD. Therefore we must choose  $x = 1$  to make  $\|\mathbf{v}_1\| = 1$ . Thus  $\mathbf{v}_1 = \begin{bmatrix} 1 & 0 \end{bmatrix}^T$ . Now for  $\mathbf{v}_2$ , we solve:

$$\begin{bmatrix} 9 - 4 & 0 \\ 0 & 4 - 4 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 5x \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

Thus, by the similar consideration, we get  $\mathbf{v}_2 = \begin{bmatrix} 0 & 1 \end{bmatrix}^T$ . Therefore we computed so far  $\Sigma = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix}$ , and  $V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ . To compute  $U$ , we use the relationship between  $\mathbf{u}_j$  and  $\mathbf{v}_j$ , i.e.,  $A \mathbf{v}_j = \sigma_j \mathbf{u}_j$ . Thus,

$$\begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = 3 \mathbf{u}_1 \implies \mathbf{u}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix},$$

$$\begin{bmatrix} 3 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = 2 \mathbf{u}_2 \implies \mathbf{u}_2 = \begin{bmatrix} 0 \\ -1 \end{bmatrix}.$$

Therefore,  $U = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ . Finally, we get:

$$A = U \Sigma V^T = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}.$$

(b)  $\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$ . First compute  $A^T A$ , which is  $\begin{bmatrix} 4 & 0 \\ 0 & 9 \end{bmatrix}$ . The characteristic equation is:

$$\det(A^T A - \lambda I) = \det\left(\begin{bmatrix} 4 - \lambda & 0 \\ 0 & 9 - \lambda \end{bmatrix}\right) = (4 - \lambda)(9 - \lambda) = 0.$$

Thus, the eigenvalues are:  $\lambda = 4, 9$ . Therefore, we immediately get the singular values of  $A$  as  $\sigma_1 = 3$ , and  $\sigma_2 = 2$ . (We must have  $\sigma_1 \geq \sigma_2$ .) Now, for each eigenvalue, we compute the corresponding eigenvector.

$$\begin{bmatrix} 4 - 9 & 0 \\ 0 & 9 - 9 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -5x \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

By the same reasoning as in (a), we have  $\mathbf{v}_1 = [0 \ 1]^T$ . Now for  $\mathbf{v}_2$ , we solve:

$$\begin{bmatrix} 4 - 4 & 0 \\ 0 & 9 - 4 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 5y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

Thus we get  $\mathbf{v}_2 = [1 \ 0]^T$ . Therefore we computed so far  $\Sigma = \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix}$ , and  $V = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ . To compute  $U$ , we again use the relationship between  $\mathbf{u}_j$  and  $\mathbf{v}_j$ , i.e.,  $A\mathbf{v}_j = \sigma_j\mathbf{u}_j$ . Thus,

$$\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = 3\mathbf{u}_1 \implies \mathbf{u}_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix},$$

$$\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = 2\mathbf{u}_2 \implies \mathbf{u}_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}.$$

Therefore,  $U = [0 \ 11 \ 0]$ . Finally, we get:

$$A = U\Sigma V^T = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}.$$

(c)  $\begin{bmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$ . First compute  $A^T A$ , which is  $\begin{bmatrix} 0 & 0 \\ 0 & 4 \end{bmatrix}$ . The characteristic equation is:

$$\det(A^T A - \lambda I) = \det\left(\begin{bmatrix} 0 - \lambda & 0 \\ 0 & 4 - \lambda \end{bmatrix}\right) = -\lambda(4 - \lambda) = 0.$$

Thus, the eigenvalues are:  $\lambda = 4, 0$ . Therefore, we immediately get the singular values of  $A$  as  $\sigma_1 = 2$ , and  $\sigma_2 = 0$  (We must choose  $\sigma_1 \geq \sigma_2$ ). Now, we compute the eigenvectors as before.

$$\begin{bmatrix} 0 - 4 & 0 \\ 0 & 4 - 4 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -4x \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

Thus  $\mathbf{v}_1 = [0 \ 1]^T$ . Now for  $\mathbf{v}_2$ , we solve:

$$\begin{bmatrix} 0-0 & 0 \\ 0 & 4-0 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 4y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

Thus, we get  $\mathbf{v}_2 = [1 \ 0]^T$ . Therefore we computed so far  $\hat{\Sigma} = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$ , and  $V = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ .

To compute  $\hat{U}$ , we use the relationship between  $\mathbf{u}_j$  and  $\mathbf{v}_j$ , i.e.,  $A\mathbf{v}_j = \sigma_j\mathbf{u}_j$ . Thus,

$$\begin{bmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = 2\mathbf{u}_1 \implies \mathbf{u}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}.$$

Now, to determine  $\mathbf{u}_2$ , the singular value  $\sigma_2 = 0$ , and

$$\begin{bmatrix} 0 & 2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = 0\mathbf{u}_2.$$

This equation is satisfied by any  $\mathbf{u}_2 \in \mathbb{R}^3$ . Now, however, we must choose  $\mathbf{u}_2$  that is perpendicular to  $\mathbf{u}_1$  and the unit length. Thus, we choose  $\mathbf{u}_2 = [0 \ 1 \ 0]^T$ . Therefore,

$\hat{U} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$ . Finally, we get the following *reduced* SVD:

$$A = \hat{U}\hat{\Sigma}V^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}.$$

If we want to compute the *full* SVD, we need to compute  $\mathbf{u}_3$ , which must be of unit norm and perpendicular to  $\mathbf{u}_1$  and  $\mathbf{u}_2$ . Thus, we can easily obtain  $\mathbf{u}_3 = [0 \ 0 \ 1]^T$ . The singular value  $\sigma_3$  must be 0 since all the singular values must be nonnegative and ordered in a non-increasing manner. Thus, we have:

$$A = U\Sigma V^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}.$$

(d)  $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ . First compute  $A^T A$ , which is  $\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ . The characteristic equation is:

$$\det(A^T A - \lambda I) = \det \left( \begin{bmatrix} 1-\lambda & 1 \\ 1 & 1-\lambda \end{bmatrix} \right) = (1-\lambda)^2 - 1^2 = \lambda(2-\lambda) = 0.$$

Thus, the eigenvalues are:  $\lambda = 2, 0$ . Therefore, we immediately get the singular values of  $A$  as  $\sigma_1 = \sqrt{2}$ , and  $\sigma_2 = 0$ . (We must have  $\sigma_1 \geq \sigma_2$ .) Now, for each eigenvalue, we compute the corresponding eigenvector.

$$\begin{bmatrix} 1-2 & 1 \\ 1 & 1-2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -x+y \\ x-y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

Thus we have the solution  $\mathbf{v}_1 = [x \ x]^T$  for arbitrary  $x \in \mathbb{R}$  in general, but we must have  $\|\mathbf{v}_1\| = 1$ . Thus we have  $\mathbf{v}_1 = [1/\sqrt{2} \ 1/\sqrt{2}]^T$ . Now for  $\mathbf{v}_2$ , we solve:

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x+y \\ x+y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

Thus we get  $\mathbf{v}_2 = [x \ -x]^T$ . Again to have the unit norm, we get:  $\mathbf{v}_2 = [1/\sqrt{2} \ -1/\sqrt{2}]^T$ . Therefore we computed so far  $\Sigma = \begin{bmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{bmatrix}$ , and  $V = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}$ . To compute  $U$ , we again use the relationship between  $\mathbf{u}_j$  and  $\mathbf{v}_j$ , i.e.,  $A\mathbf{v}_j = \sigma_j\mathbf{u}_j$ . Thus,

$$\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = \sqrt{2}\mathbf{u}_1 \implies \mathbf{u}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}.$$

For  $\mathbf{u}_2$ , because  $\sigma_2 = 0$ , we can choose the unit vector perpendicular to  $\mathbf{u}_1$ . Thus, we get  $\mathbf{u}_2 = [0 \ 1]^T$ . Therefore,  $U = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ . Finally, we get:

$$A = U\Sigma V^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}.$$

(e)  $\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ . First compute  $A^T A$ , which is  $\begin{bmatrix} 2 & 2 \\ 2 & 2 \end{bmatrix}$ . The characteristic equation is:

$$\det(A^T A - \lambda I) = \det\left(\begin{bmatrix} 2-\lambda & 2 \\ 2 & 2-\lambda \end{bmatrix}\right) = (2-\lambda)^2 - 2^2 = \lambda(4-\lambda) = 0.$$

Thus, the eigenvalues are:  $\lambda = 4, 0$ . Therefore, we immediately get the singular values of  $A$  as  $\sigma_1 = 2$ , and  $\sigma_2 = 0$ . (We must have  $\sigma_1 \geq \sigma_2$ .) Now, for each eigenvalue, we compute the corresponding eigenvector.

$$\begin{bmatrix} 2-4 & 2 \\ 2 & 2-4 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -2x+2y \\ 2x-2y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

Thus we have the solution  $\mathbf{v}_1 = [x \ x]^T$  for arbitrary  $x \in \mathbb{R}$  in general, but we must have  $\|\mathbf{v}_1\| = 1$ . Thus we have  $\mathbf{v}_1 = [1/\sqrt{2} \ 1/\sqrt{2}]^T$ . Now for  $\mathbf{v}_2$ , we solve:

$$\begin{bmatrix} 2 & 2 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 2x + 2y \\ 2x + 2y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

Thus we get  $\mathbf{v}_2 = [x \ -x]^T$ . Again to have the unit norm, we get:  $\mathbf{v}_2 = [1/\sqrt{2} \ -1/\sqrt{2}]^T$ . Therefore we computed so far  $\Sigma = \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}$ , and  $V = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}$ . To compute  $U$ , we again use the relationship between  $\mathbf{u}_j$  and  $\mathbf{v}_j$ , i.e.,  $A\mathbf{v}_j = \sigma_j\mathbf{u}_j$ . Thus,

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix} = 2\mathbf{u}_1 \implies \mathbf{u}_1 = [1/\sqrt{2} \ 1/\sqrt{2}].$$

For  $\mathbf{u}_2$ , because  $\sigma_2 = 0$ , we can choose the unit vector perpendicular to  $\mathbf{u}_1$ . Thus, we can get, for example,  $\mathbf{u}_2 = [1/\sqrt{2} \ -1/\sqrt{2}]^T$ . Therefore,  $U = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}$ . Finally, we get:

$$A = U\Sigma V^T = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}.$$

**Exercise 4.4** This is *false*. The best explanation is to consider the following counterexample. Let

$A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ , and  $B = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ . It is easy to see that the singular values of  $A$  and  $B$  are the same, i.e.,  $\sigma_1 = 1, \sigma_2 = 0$ . (Do the similar computation as in Exercise 4.1.) The question is, therefore, where there is any unitary matrix  $Q$  that satisfies  $A = QBQ^*$ , in other words,  $AQ = QB$ . OK, so let's  $Q = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ . Then,

$$AQ = \begin{bmatrix} a & b \\ 0 & 0 \end{bmatrix} = QB = \begin{bmatrix} 0 & a \\ 0 & c \end{bmatrix}.$$

Thus we must have  $a = b = c = 0$ . Now,  $Q = \begin{bmatrix} 0 & 0 \\ 0 & d \end{bmatrix}$ , which cannot be unitary, i.e.,  $Q^*Q = I$ , regardless of the values of  $d$ . Thus the statement of this exercise is false.

**Exercise 4.5** The easiest way to prove this statement is to use the following fact:

**Theorem:** Let  $B \in \mathbb{R}^{n \times n}$ . Then,  $B$  is a real symmetric matrix if and only if  $B$  has a real eigenvalue-eigenvector decomposition, i.e.,  $B = P\Lambda P^T$ , where  $P$  is an  $n \times n$  orthogonal matrix and  $\Lambda$  is an  $n \times n$  real-valued diagonal matrix.

If we decide to use the above theorem, it is easy to prove this exercise. For any  $A \in \mathbb{R}^{m \times n}$ , it has a SVD,  $A = U\Sigma V^*$ . At this point, we do not know whether  $U$  and  $V$  are real-valued matrix. The only thing we know is that they are unitary. Now,  $A^T A$  is an  $n \times n$  real symmetric matrix. Thus,  $A^T A$  must have real-valued eigenvalues and eigenvectors. Because the eigenvalues of  $A^T A$  is the square of the singular values of  $A$ , we have  $A^T A = V\Sigma^T \Sigma V^*$  where  $V$  plays the role of  $P$ , and  $\Sigma^T \Sigma$  plays the role of  $\Lambda$  in the above theorem. Therefore,  $V$  must be a real-valued orthogonal matrix. Now, consider  $AA^T$  and again this is a real symmetric matrix. With the same reasoning,  $U$  must be a real-valued orthogonal matrix.

Another possible approach to prove this problem is to follow the proof of Theorem 4.1 in the Lecture 4 note by replace every occurrence of  $\mathbb{C}$  and “unitary” to  $\mathbb{R}$  and “orthogonal”.

**Exercise 5.1** In this problem, as I announced in my email, the matrix  $A$  is  $\begin{bmatrix} 1 & 2 \\ 0 & 2 \end{bmatrix}$ . We just follow

the same strategy as Exercise 4.1. Since  $A^T A = \begin{bmatrix} 1 & 2 \\ 2 & 8 \end{bmatrix}$ , the characteristic equation is:

$$\det(A^T A - \lambda I) = \det \left( \begin{bmatrix} 1 - \lambda & 2 \\ 2 & 8 - \lambda \end{bmatrix} \right) = (1 - \lambda)(8 - \lambda) - 2^2 = \lambda^2 - 9\lambda + 4 = 0.$$

Thus  $\lambda = \frac{9 \pm \sqrt{65}}{2}$ . Now, the singular values are:  $\sigma_1 = \sqrt{\frac{9 + \sqrt{65}}{2}} \approx 2.9208$ , and  $\sigma_2 = \sqrt{\frac{9 - \sqrt{65}}{2}} \approx 0.6847$ .

Just in case, the eigenvectors of  $A^T A$  are:

$$\begin{bmatrix} 1 - \frac{9 + \sqrt{65}}{2} & 2 \\ 2 & 8 - \frac{9 + \sqrt{65}}{2} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -\frac{7 + \sqrt{65}}{2}x + 2y \\ 2x + \frac{7 - \sqrt{65}}{2}y \end{bmatrix} = \mathbf{0}.$$

From here we get  $\frac{7 + \sqrt{65}}{2}x = 2y$ . Then, setting, say  $x = 4$ ,  $y = 7 + \sqrt{65}$ . But  $\mathbf{v}_1$  must be of the unit norm. Thus, by normalizing it to get:

$$\mathbf{v}_1 = \left[ \sqrt{\frac{8}{65 + 7\sqrt{65}}} \quad \sqrt{\frac{57 + 7\sqrt{65}}{65 + 7\sqrt{65}}} \right]^T.$$

Similarly, for  $\mathbf{v}_2$ , we have:

$$\begin{bmatrix} 1 - \frac{9 - \sqrt{65}}{2} & 2 \\ 2 & 8 - \frac{9 - \sqrt{65}}{2} \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -\frac{7 - \sqrt{65}}{2}x + 2y \\ 2x + \frac{7 + \sqrt{65}}{2}y \end{bmatrix} = \mathbf{0}.$$

From here we get  $\frac{7 - \sqrt{65}}{2}x = 2y$ . setting, say  $x = -4$ ,  $y = -7 + \sqrt{65}$ . But  $\mathbf{v}_1$  must be of the unit norm. Thus, by normalizing it to get:

$$\mathbf{v}_2 = \left[ -\sqrt{\frac{8}{65 - 7\sqrt{65}}} \quad \sqrt{\frac{57 - 7\sqrt{65}}{65 - 7\sqrt{65}}} \right]^T.$$

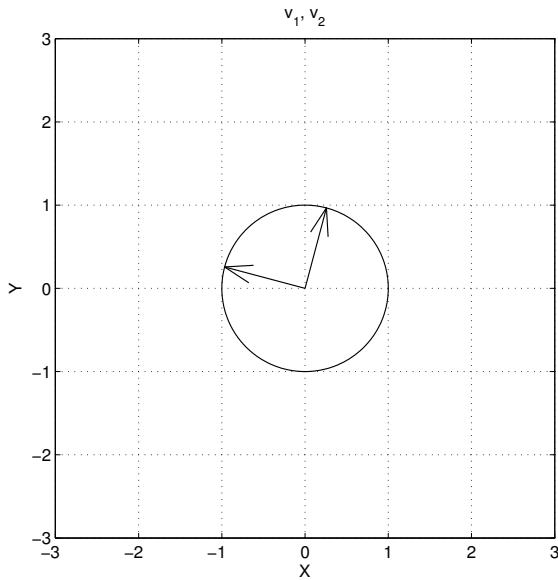
Now, we can compute  $U$  via:

$$\mathbf{u}_1 = \frac{1}{\sigma_1} A \mathbf{v}_1 = \sqrt{\frac{2}{9 + \sqrt{65}}} \begin{bmatrix} \frac{\sqrt{8} + 2\sqrt{57 + 7\sqrt{65}}}{\sqrt{65 + 7\sqrt{65}}} \\ \frac{2\sqrt{57 + 7\sqrt{65}}}{\sqrt{65 + 7\sqrt{65}}} \end{bmatrix} = \frac{\sqrt{2}}{2\sqrt{65 + 8\sqrt{65}}} \begin{bmatrix} \sqrt{2} + \sqrt{57 + 7\sqrt{65}} \\ \sqrt{57 + 7\sqrt{65}} \end{bmatrix}.$$

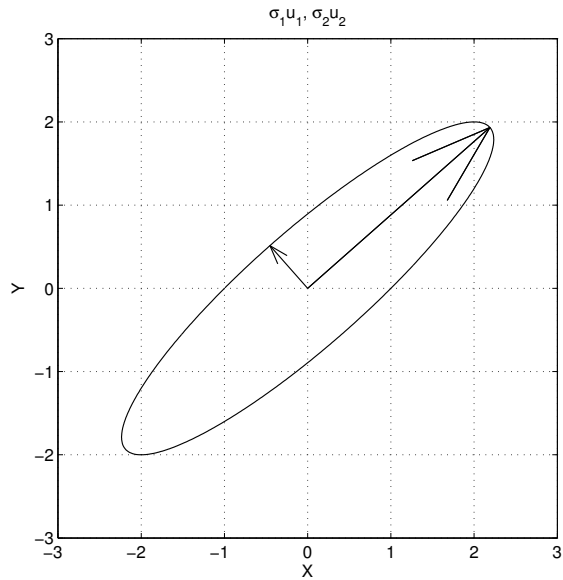
$$\mathbf{u}_2 = \frac{1}{\sigma_2} A \mathbf{v}_2 = \sqrt{\frac{2}{9 - \sqrt{65}}} \begin{bmatrix} \frac{-\sqrt{8} + 2\sqrt{57 - 7\sqrt{65}}}{\sqrt{65 + 7\sqrt{65}}} \\ \frac{2\sqrt{57 - 7\sqrt{65}}}{\sqrt{65 - 7\sqrt{65}}} \end{bmatrix} = \frac{\sqrt{2}}{2\sqrt{65 - 8\sqrt{65}}} \begin{bmatrix} -\sqrt{2} + \sqrt{57 - 7\sqrt{65}} \\ \sqrt{57 - 7\sqrt{65}} \end{bmatrix}.$$

The following figure shows how the unit sphere (circle) in  $\mathbb{R}^2$  is mapped to the ellipse by this matrix  $A$ .





(a) Unit Circle S



(b) AS

**Exercise 5.3**

$$A = \begin{bmatrix} -2 & 11 \\ -10 & 5 \end{bmatrix}$$

(a) As before, we first compute the eigenvalues and eigenvectors of  $A^T A$ .

$$A^T A = \begin{bmatrix} -2 & -10 \\ 11 & 5 \end{bmatrix} \begin{bmatrix} -2 & 11 \\ -10 & 5 \end{bmatrix} = \begin{bmatrix} 104 & -72 \\ -72 & 146 \end{bmatrix}.$$

Thus, the characteristic equation is:

$$\det(A^T A - \lambda I) = \det \begin{bmatrix} 104 - \lambda & -72 \\ -72 & 146 - \lambda \end{bmatrix} = \lambda^2 - 250\lambda + 10000 = 0,$$

$$\implies (\lambda - 200)(\lambda - 50) = 0 \implies \lambda = 200, 50 \implies \sigma_1 = 10\sqrt{2}, \sigma_2 = 5\sqrt{2}.$$

Now let's compute the eigenvectors. For the first eigenvector  $\mathbf{v}_1 = [x \ y]^T$ , we need to solve  $A^T A \mathbf{v}_1 = 200\mathbf{v}_1$ :

$$\begin{bmatrix} 104x - 72y \\ -72x + 146y \end{bmatrix} = \begin{bmatrix} 200x \\ 200y \end{bmatrix}$$

Thus,

$$\begin{aligned} 96x &= -72y \\ -72x &= 54y, \end{aligned}$$

which are the same equation:  $4x = -3y$ . Therefore, we can take, say,  $x = -3; y = 4$ . Note that in this case, it is impossible for  $x$  and  $y$  to have the same sign. Since  $\|\mathbf{v}_1\| = 1$ , we normalize it to have  $\mathbf{v}_1 = [-3/5 \ 4/5]$ . (Of course, it is also possible to have  $\mathbf{v}_1 = [3/5 \ -4/5]$ . This choice is up to you. Both are correct.) As for the second eigenvector  $\mathbf{v}_2$ , we have to solve:

$$\begin{bmatrix} 104x - 72y \\ -72x + 146y \end{bmatrix} = \begin{bmatrix} 50x \\ 50y \end{bmatrix}$$

Thus,

$$\begin{aligned} 54x &= 72y \\ 3x &= 4y, \end{aligned}$$

which are the same equation:  $3x = 4y$ . In this case, we can take say,  $x = 4; y = 3$ , of the same sign. After the normalization, we have  $\mathbf{v}_2 = [4/5 \ 3/5]$ .

So far, we computed:

$$\begin{aligned} \Sigma &= \begin{bmatrix} 10\sqrt{2} & 0 \\ 0 & 5\sqrt{2} \end{bmatrix}, \\ V &= \begin{bmatrix} -\frac{3}{5} & \frac{4}{5} \\ \frac{4}{5} & \frac{3}{5} \end{bmatrix}. \end{aligned}$$

As for  $U$ , we use the formula  $\mathbf{u}_j = \frac{1}{\sigma_j} A \mathbf{v}_j$ ,  $j = 1, 2$ :

$$\mathbf{u}_1 = \frac{1}{10\sqrt{2}} \begin{bmatrix} -2 & 11 \\ -10 & 5 \end{bmatrix} \begin{bmatrix} -3/5 \\ 4/5 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}.$$

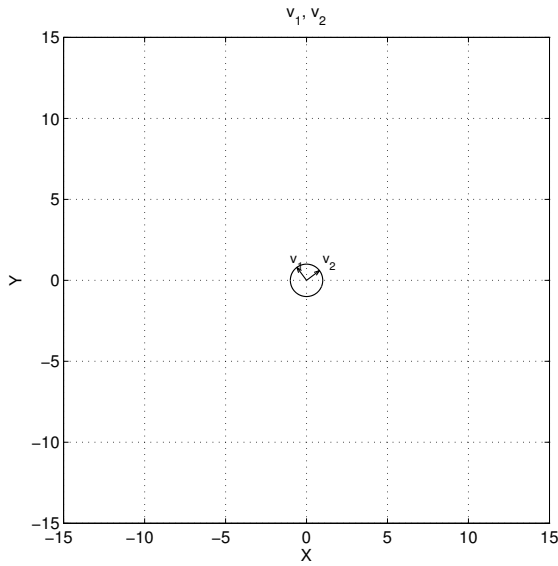
$$\mathbf{u}_2 = \frac{1}{5\sqrt{2}} \begin{bmatrix} -2 & 11 \\ -10 & 5 \end{bmatrix} \begin{bmatrix} 4/5 \\ 3/5 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}.$$

Therefore we have:

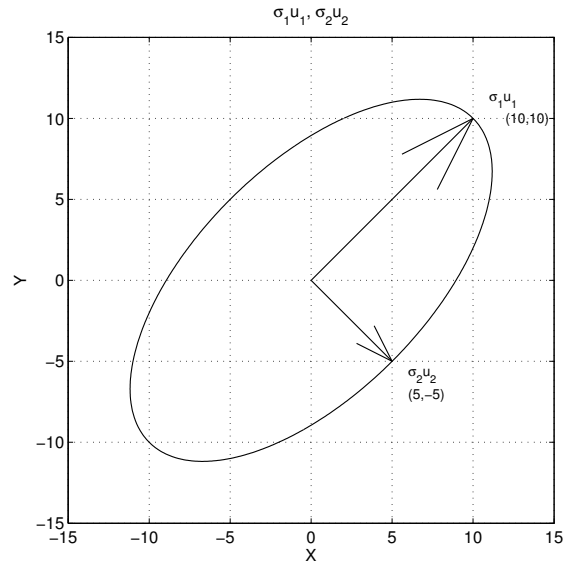
$$U = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix}.$$

Finally, we have the following SVD of  $A$ :

$$A = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} 10\sqrt{2} & 0 \\ 0 & 5\sqrt{2} \end{bmatrix} \begin{bmatrix} -\frac{3}{5} & \frac{4}{5} \\ \frac{4}{5} & \frac{3}{5} \end{bmatrix}.$$



(b.1) Unit Circle S



(b.2) AS

(b) See the above figures.

(c) The 1-norm of an matrix  $A$  is the largest 1-norm of the column vectors of  $A$ . Thus,

$$\|A\|_1 = \max_{j=1,2} \|A(:, j)\|_1 = 16.$$

The 2-norm of  $A$  is of course the largest singular value  $\sigma_1$ . Thus,

$$\|A\|_2 = \sigma_1 = 10\sqrt{2}.$$

The  $\infty$ -norm of  $A$  is the largest 1-norm of the *row* vectors of  $A$ . Thus,

$$\|A\|_\infty = \max_{i=1,2} \|A(i, :)\|_1 = 15.$$

(d) In this case, the diagonal elements of  $\Sigma$  are not zeros, so we can compute the exact inverse of  $A$ :

$$\begin{aligned} A^{-1} &= V\Sigma^{-1}U^* = \begin{bmatrix} -\frac{3}{5} & \frac{4}{5} \\ \frac{4}{5} & \frac{3}{5} \end{bmatrix} \begin{bmatrix} \frac{1}{10\sqrt{2}} & 0 \\ 0 & \frac{1}{5\sqrt{2}} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} \\ &= \begin{bmatrix} -\frac{3}{5} & \frac{4}{5} \\ \frac{4}{5} & \frac{3}{5} \end{bmatrix} \begin{bmatrix} \frac{1}{10} & 0 \\ 0 & \frac{1}{5} \end{bmatrix} \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{bmatrix} = \frac{1}{100} \begin{bmatrix} 5 & -11 \\ 10 & -2 \end{bmatrix}. \end{aligned}$$

(e)

$$\det(A - \lambda I) = \det \begin{bmatrix} -2 - \lambda & 11 \\ -10 & 5 - \lambda \end{bmatrix} = (-2 - \lambda)(5 - \lambda) + 100 = 0$$

$$\implies (\lambda - 5)(\lambda + 2) + 100 = 0 \implies \lambda^2 - 3\lambda + 100 = 0.$$

This leads to

$$\lambda = \frac{3 \pm \sqrt{391}i}{2},$$

i.e.,

$$\lambda_1 = \frac{3 + \sqrt{391}i}{2}, \quad \lambda_2 = \frac{3 - \sqrt{391}i}{2}.$$

(f) This is just a simple computation.

$$\det(A) = -2 * 5 + 110 = 100.$$

$$\lambda_1 \lambda_2 = \frac{9 + 391}{4} = 100.$$

$$|\det(A)| = \sigma_1 \sigma_2 = 10\sqrt{2} \cdot 5\sqrt{2} = 100.$$

(g) The area of the ellipsoid is  $\pi$  times the length of the major axis times the length of the minor axis. Thus, it is  $\pi \sigma_1 \sigma_2 = 100\pi$ .