# Homework 5 Solutions

## Math 198: Math for Machine Learning

Due Date: March 11

Name: Student ID:

## 1 Working with Adjoints

Let  $\mathbf{A} \in \mathbb{R}^{m \times n}$ .

(a) Show that  $\ker \mathbf{A}^{\top} \mathbf{A} = \ker \mathbf{A}$ .

$$\mathbf{v} \in \ker(\mathbf{A}^{\top} \mathbf{A}) \iff \mathbf{A}^{\top} \mathbf{A} \mathbf{v} = 0$$

$$\iff \mathbf{A} \mathbf{v} \in \ker(\mathbf{A}^{\top})$$

$$\iff \mathbf{A} \mathbf{v} \in \operatorname{range}(\mathbf{A})^{\perp} \text{ by FTLA}$$

$$\iff \mathbf{A} \mathbf{v} = \mathbf{0} \text{ since } \mathbf{A} \mathbf{v} \in \operatorname{range}(\mathbf{A})$$

$$\iff \mathbf{v} \in \ker(\mathbf{A})$$

Therefore,  $\ker(\mathbf{A}^{\top}\mathbf{A}) = \ker(\mathbf{A})$ .

- (b) Deduce that  $\operatorname{rank}(\mathbf{A}^{\top}\mathbf{A}) = \operatorname{rank}(\mathbf{A})$ . We have that  $\operatorname{rank}(\mathbf{A}) + \dim \ker(\mathbf{A}) = n = \operatorname{rank}(\mathbf{A}^{\top}\mathbf{A}) + \dim \ker(\mathbf{A}^{\top}\mathbf{A})$ . Since  $\dim \ker(\mathbf{A}) = \dim \ker(\mathbf{A}^{\top}\mathbf{A})$ ,  $\operatorname{rank}(\mathbf{A}) = \operatorname{rank}(\mathbf{A}^{\top}\mathbf{A})$ .
- (c) Suppose **A** is square. Show that **A** and  $\mathbf{A}^{\top}$  have the same eigenvalues.

$$\begin{aligned} p_{\mathbf{A}}(\lambda) &= \det(\mathbf{A} - \lambda \mathbf{I}) \\ &= \det(\mathbf{A} - \lambda \mathbf{I})^{\top} \\ &= \det(\mathbf{A}^{\top} - \lambda \mathbf{I}) \\ &= p_{\mathbf{A}^{\top}}(\lambda) \end{aligned}$$

as determinant is invariant under transpose. Since the two matrices have the same characteristic polynomial, they have the same eigenvalues.

(d) Deduce that  $\operatorname{rank}(\mathbf{A}) = \operatorname{rank}(\mathbf{A}^{\top})$ . Because the matrices have the same eigenvalues, they have the same number of non-zero eigenvalues. Therefore, they have the same rank.

#### 2 SVD

(a) Let  $\mathbf{A} \in \mathbb{R}^{m \times n}$ . Let  $\mathbf{A} = \mathbf{U} \Sigma \mathbf{V}^{\top}$  be its SVD. Find the spectral decompositions of  $\mathbf{A}^{\top} \mathbf{A}$  and  $\mathbf{A} \mathbf{A}^{\top}$  in terms of  $\mathbf{U}, \Sigma, \mathbf{V}$ .

$$\mathbf{A}^{\top} \mathbf{A} = (\mathbf{U} \mathbf{\Sigma} \mathbf{V}^{\top})^{\top} \mathbf{U} \mathbf{\Sigma} \mathbf{V}^{\top}$$
$$= \mathbf{V} \mathbf{\Sigma} \mathbf{U} \mathbf{U}^{\top} \mathbf{\Sigma} \mathbf{V}^{\top}$$
$$= \mathbf{V} \mathbf{\Sigma}^{2} \mathbf{V}^{\top}$$

$$\begin{aligned} \mathbf{A}\mathbf{A}^\top &= \mathbf{U}\boldsymbol{\Sigma}\mathbf{V}^\top(\mathbf{U}\boldsymbol{\Sigma}\mathbf{V}^\top)^\top \\ &= \mathbf{U}\boldsymbol{\Sigma}\mathbf{V}^\top\mathbf{V}\boldsymbol{\Sigma}\mathbf{U}^\top \\ &= \mathbf{U}\boldsymbol{\Sigma}\mathbf{U}^\top \end{aligned}$$

- (b) Prove: If  $\mathbf{A} \in \mathbb{R}^{n \times n}$  is PSD, then the spectral decomposition of  $\mathbf{A}$  coincides with the SVD of  $\mathbf{A}$ . (The technical details involving rows or columns of zeros are omitted.) Let  $\mathbf{A} = \mathbf{U} \boldsymbol{\Sigma} \mathbf{V}^{\top}$  by SVD, and  $\mathbf{A} = \mathbf{U}' \boldsymbol{\Lambda} \mathbf{U}'^{\top}$  by the Spectral Theorem. We seek to show that  $\mathbf{U} = \mathbf{U}' = \mathbf{V}$  and  $\boldsymbol{\Sigma} = \boldsymbol{\Lambda}$ . Recall that  $\mathbf{V}$  is the unitary matrix in the spectral decomposition  $\mathbf{A}^{\top} \mathbf{A} = \mathbf{V} \boldsymbol{\Sigma}^{2} \mathbf{V}^{\top}$ . But  $\mathbf{A}^{\top} \mathbf{A} = (\mathbf{U}' \boldsymbol{\Lambda} \mathbf{U}'^{\top})^{\top} \mathbf{U}' \boldsymbol{\Lambda} \mathbf{U}'^{\top} = \mathbf{U}' \boldsymbol{\Lambda}^{2} \mathbf{U}'^{\top}$ . So  $\mathbf{V} = \mathbf{U}'$ , and since  $\boldsymbol{\Lambda}$  and  $\boldsymbol{\Sigma}$  are PSD,  $\boldsymbol{\Lambda} = \boldsymbol{\Sigma}$ . So  $\mathbf{U} \boldsymbol{\Sigma} \mathbf{V}^{\top} = \mathbf{U}' \boldsymbol{\Sigma} \mathbf{V}^{\top}$ , and so  $\mathbf{U} = \mathbf{U}'$ .
- (c) Let  $\mathbf{A} \in \mathbb{R}^{m \times n}$ , and let  $\mathbf{A} = \mathbf{U} \Sigma \mathbf{V}^{\top}$  be its SVD. If  $\mathbf{V} = (\mathbf{v}_1, \dots, \mathbf{v}_n)$ ,  $\mathbf{U} = (\mathbf{u}_1, \dots, \mathbf{u}_m)$ , and  $r = \operatorname{rank}(\mathbf{A})$ , then let

$$\mathbf{V}_r = (\mathbf{v}_1, \dots, \mathbf{v}_r)$$
  
 $\mathbf{U}_r = (\mathbf{u}_1, \dots, \mathbf{u}_r).$ 

Show that  $\mathbf{v}_1, \dots, \mathbf{v}_r$  "diagonalize"  $\mathbf{A}$  in the following way: For  $i = 1, \dots, r$ , show that  $\mathbf{A}\mathbf{v}_i = \sigma_i \mathbf{u}_i$ . Because  $\mathbf{V}$  is unitary,  $\mathbf{V}^{\top}\mathbf{v}_i$  will be 0 in every index except  $(\mathbf{V}\mathbf{v}_i)_i = 1$ . Since  $\mathbf{A}\mathbf{v}_i = \mathbf{U}\boldsymbol{\Sigma}\mathbf{V}^{\top}\mathbf{v}_i$ ,

$$\mathbf{A}\mathbf{v}_i = \mathbf{U} \begin{bmatrix} 0 & \dots & \sigma_i & \dots & 0 \end{bmatrix}^{\top} = \sigma_i \mathbf{u}_i$$

(d) Let

$$\mathbf{A} = \begin{pmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{pmatrix}.$$

Compute the SVD of  $\mathbf{A}$ . We first compute  $\mathbf{A}^{\top}\mathbf{A}$ :

$$\mathbf{A}^{\top}\mathbf{A} = \begin{bmatrix} 3 & 2 \\ 2 & 3 \\ 2 & -2 \end{bmatrix} \begin{bmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{bmatrix} = \begin{bmatrix} 13 & 12 & 2 \\ 12 & 13 & -2 \\ 2 & -2 & 8 \end{bmatrix}$$

Next, we find the eigenvalues of  $\mathbf{A}^{\top}\mathbf{A}$ :

$$\begin{split} p_{\mathbf{A}^{\top}\mathbf{A}}(\lambda) &= \det \begin{bmatrix} 13 - \lambda & 12 & 2 \\ 12 & 13 - \lambda & -2 \\ 2 & -2 & 8 - \lambda \end{bmatrix} \\ &= (13 - \lambda) \det \begin{bmatrix} 13 - \lambda & -2 \\ -2 & 8 - \lambda \end{bmatrix} - 12 \det \begin{bmatrix} 12 & -2 \\ 2 & 8 - \lambda \end{bmatrix} + 2 \det \begin{bmatrix} 12 & 13 - \lambda \\ 2 & -2 \end{bmatrix} \\ &= (13 - \lambda)((13 - \lambda)(8 - \lambda) - 4) - 12(12(8 - \lambda) + 4) + 2(-24 - 2(13 - \lambda)) \\ &= (13 - \lambda)(100 - 21\lambda + \lambda^2) - 12(100 - 12\lambda) + 2(-50 + 2\lambda) \\ &= -\lambda^3 + 34\lambda^2 - 225\lambda \end{split}$$

It is immediately clear that one eigenvalue is 0, as expected, since  $rank(\mathbf{A}) = 2$ . To find the other two, we factor:

$$-\lambda^2 + 34\lambda - 225 = -(\lambda - 9)(\lambda - 25)$$

So the eigenvalues of  $\mathbf{A}^{\top}\mathbf{A}$  are 0, 9, and 25. We now find the eigenvectors of  $\mathbf{A}^{\top}\mathbf{A}$ . We first find the eigenvector corresponding to  $\lambda_1 = 25$ :

$$\mathbf{v}_1 \in \ker(\mathbf{A}^{\top} \mathbf{A} - 25\mathbf{I})$$

$$\in \ker \begin{bmatrix} -12 & 12 & 2 \\ 12 & -12 & -2 \\ 2 & -2 & -17 \end{bmatrix}$$

Observe that  $(\mathbf{A}^{\top}\mathbf{A} - 25\mathbf{I})_1 = -(\mathbf{A}^{\top}\mathbf{A} - 25\mathbf{I})_2$ . Therefore,  $[1\ 1\ 0]^{\top}$  is an eigenvector of  $\mathbf{A}$  corresponding to  $\lambda_1 = 25$ . We turn next to  $\lambda_2 = 9$ :

$$\mathbf{v}_2 \in \ker(\mathbf{A}^{\top} \mathbf{A} - 9\mathbf{I})$$

$$\in \ker \begin{bmatrix} 4 & 12 & 2 \\ 12 & 4 & -2 \\ 2 & -2 & -1 \end{bmatrix}$$

Observe that  $(\mathbf{A}^{\top}\mathbf{A} - 9\mathbf{I})_2 = (\mathbf{A}^{\top}\mathbf{A} - 9\mathbf{I})_1 + 4(\mathbf{A}^{\top}\mathbf{A} - 9\mathbf{I})_3$ . Therefore,  $[1 \ -1 \ 4]^{\top}$  is an eigenvector of  $\mathbf{A}$  corresponding to  $\lambda_2 = 9$ . We turn next to  $\lambda_3 = 0$ :

$$\mathbf{v}_3 \in \ker(\mathbf{A}^{\top} \mathbf{A})$$

$$\in \ker \begin{bmatrix} 13 & 12 & 2 \\ 12 & 13 & -2 \\ 2 & -2 & 8 \end{bmatrix}$$

Observe that  $2(\mathbf{A}^{\top}\mathbf{A})_1 - 2(\mathbf{A}^{\top}\mathbf{A})_2 = (\mathbf{A}^{\top}\mathbf{A})_3$ . Therefore,  $[2 - 2 - 1]^{\top}$  is an eigenvector of  $\mathbf{A}$  corresponding to  $\lambda_3 = 0$ . Note that the eigenvectors we have collected are already an orthogonal basis for  $\mathbb{R}^3$ ; we just need to normalize them to find the column vectors of  $\mathbf{V}$ . Therefore,

$$\mathbf{V} = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{3\sqrt{2}} & \frac{2}{3} \\ \frac{1}{\sqrt{2}} & -\frac{1}{3\sqrt{2}} & -\frac{2}{3} \\ 0 & \frac{4}{3\sqrt{2}} & -\frac{1}{3} \end{bmatrix}, \mathbf{\Lambda} = \begin{bmatrix} 25 & 0 & 0 \\ 0 & 9 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \mathbf{A}^{\top} \mathbf{A} = \mathbf{V} \mathbf{\Lambda} \mathbf{V}^{\top}$$

So, for our SVD, we have

$$\Sigma = \begin{bmatrix} 5 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \mathbf{V} = \mathbf{V}$$
 from spectral decomposition

We can then solve for U:

$$\mathbf{U}' = \mathbf{A}\mathbf{V}\mathbf{\Sigma}^{-1}$$

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

$$= \begin{bmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{bmatrix} \begin{bmatrix} \frac{1}{5\sqrt{2}} & \frac{1}{9\sqrt{2}} & 0 \\ \frac{1}{5\sqrt{2}} & -\frac{1}{9\sqrt{2}} & 0 \\ 0 & \frac{4}{9\sqrt{2}} & 0 \end{bmatrix}$$

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

So the SVD of A is

$$\mathbf{A} = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0\\ \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} & 0 \end{bmatrix} \begin{bmatrix} 5 & 0 & 0\\ 0 & 3 & 0\\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{3\sqrt{2}} & \frac{2}{3}\\ \frac{1}{\sqrt{2}} & -\frac{1}{3\sqrt{2}} & -\frac{2}{3}\\ 0 & \frac{4}{3\sqrt{2}} & -\frac{1}{3} \end{bmatrix}^{\top}$$

$$\mathbf{A} = \begin{pmatrix} 3 & 2 & 2 \\ 2 & 3 & -2 \end{pmatrix}.$$

Find orthonormal bases for the four fundamental subspaces of **A**. From FTLA part II and the SVD from (d), we have

$$\begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{bmatrix}, \begin{bmatrix} \frac{1}{3\sqrt{2}} \\ -\frac{1}{3\sqrt{2}} \\ \frac{4}{3\sqrt{2}} \end{bmatrix} \text{ are a basis for range}(\mathbf{A}^{\top})$$

$$\begin{bmatrix} \frac{2}{3} \\ -\frac{2}{3} \\ -\frac{1}{3} \end{bmatrix}$$
 is a basis for  $\ker(\mathbf{A})$ 

$$\begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix}, \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix} \text{ are a basis for range}(\mathbf{A})$$

 $\mathbf{A}^{\top}$  has a trivial kernel

## 3 PCA

Let

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

Note that the SVD of X is

$$SVD(\mathbf{X}) = \begin{bmatrix} 0 & 0.7 & -0.5 & 0 & 0.5 \\ 0 & -0.7 & -0.5 & 0 & 0.5 \\ -0.7 & 0 & 0.5 & 0 & 0.5 \\ 0 & 0 & 0 & 1 & 0 \\ 0.7 & 0 & 0.5 & 0 & 0.5 \end{bmatrix} \begin{bmatrix} 4 & 0 \\ 0 & 2 \\ 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0.7 & 0.7 \\ -0.7 & 0.7 \end{bmatrix}$$

Compute the principal components of X.

The principal components of  $\mathbf{X}$  are unit eigenvectors of  $\mathbf{X}^{\top}\mathbf{X}$ . In this case,

$$\mathbf{v}_1 = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix}, \mathbf{v}_2 = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix}$$