

Solution to Homework Assignment No. 6

1. (a) No. Let $\mathbf{v} = (1, 0)$ and $\mathbf{w} = (0, 1)$. Since

$$\begin{aligned} T(\mathbf{v}) = (1, 0), \quad T(\mathbf{w}) = (0, 1) &\implies T(\mathbf{v}) + T(\mathbf{w}) = (1, 1) \\ \mathbf{v} + \mathbf{w} = (1, 1) &\implies T(\mathbf{v} + \mathbf{w}) = (1/\sqrt{2}, 1/\sqrt{2}) \end{aligned}$$

we have $T(\mathbf{v}) + T(\mathbf{w}) \neq T(\mathbf{v} + \mathbf{w})$ and T is not a linear transformation.

- (b) Yes. Let $\mathbf{v} = (v_1, v_2, v_3)$ and $\mathbf{w} = (w_1, w_2, w_3)$. Then

$$\begin{aligned} &T(c\mathbf{v} + d\mathbf{w}) \\ &= T(cv_1 + dw_1, cv_2 + dw_2, cv_3 + dw_3) \\ &= (cv_1 + dw_1, 2cv_2 + 2dw_2, 3cv_3 + 3dw_3) \\ &= c(v_1, 2v_2, 3v_3) + d(w_1, 2w_2, 3w_3) \\ &= cT(\mathbf{v}) + dT(\mathbf{w}). \end{aligned}$$

Hence, T is a linear transformation.

- (c) No. Let $\mathbf{v} = (1, 0)$ and $\mathbf{w} = (0, 1)$. Since

$$\begin{aligned} T(\mathbf{v}) = 1, \quad T(\mathbf{w}) = 1 &\implies T(\mathbf{v}) + T(\mathbf{w}) = 2 \\ \mathbf{v} + \mathbf{w} = (1, 1) &\implies T(\mathbf{v} + \mathbf{w}) = 1 \end{aligned}$$

we have $T(\mathbf{v}) + T(\mathbf{w}) \neq T(\mathbf{v} + \mathbf{w})$ and T is not a linear transformation.

2. (a) Let $\mathbf{X}, \mathbf{Y} \in M$. Then we can have

$$T(a\mathbf{X} + b\mathbf{Y}) = \mathbf{A}(a\mathbf{X} + b\mathbf{Y}) = a\mathbf{A}\mathbf{X} + b\mathbf{A}\mathbf{Y} = aT(\mathbf{X}) + bT(\mathbf{Y})$$

and hence T is linear.

- (b) Since $\beta = \{\mathbf{V}_1, \mathbf{V}_2, \mathbf{V}_3, \mathbf{V}_4\}$ and

$$\begin{aligned} T(\mathbf{V}_1) &= \begin{bmatrix} a & 0 \\ c & 0 \end{bmatrix} = a\mathbf{V}_1 + c\mathbf{V}_3 \\ T(\mathbf{V}_2) &= \begin{bmatrix} 0 & a \\ 0 & c \end{bmatrix} = a\mathbf{V}_2 + c\mathbf{V}_4 \\ T(\mathbf{V}_3) &= \begin{bmatrix} b & 0 \\ d & 0 \end{bmatrix} = b\mathbf{V}_1 + d\mathbf{V}_3 \\ T(\mathbf{V}_4) &= \begin{bmatrix} 0 & b \\ 0 & d \end{bmatrix} = b\mathbf{V}_2 + d\mathbf{V}_4 \end{aligned}$$

we can have

$$[T]_{\beta} = \begin{bmatrix} a & 0 & b & 0 \\ 0 & a & 0 & b \\ c & 0 & d & 0 \\ 0 & c & 0 & d \end{bmatrix}.$$

3. (a) Since $\beta = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$, $\gamma = \{\mathbf{w}_1, \mathbf{w}_2, \mathbf{w}_3\}$, and

$$\begin{aligned} T(\mathbf{v}_1) &= \mathbf{w}_2 \\ T(\mathbf{v}_2) &= \mathbf{w}_1 + \mathbf{w}_3 \\ T(\mathbf{v}_3) &= \mathbf{w}_1 + \mathbf{w}_3 \end{aligned}$$

we can obtain

$$[T]_{\beta}^{\gamma} = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 1 \end{bmatrix}.$$

(b) From (a), we can find that $\{(0, 1, -1)^T\}$ forms a basis for $\mathcal{N}([T]_{\beta}^{\gamma})$. Therefore, the kernel of T is given by the span of $\mathbf{v}_2 - \mathbf{v}_3$.

(c) The dimension of the range of T is equal to that of the column space of $[T]_{\beta}^{\gamma}$, which is 2.

4. Assume $T : V \rightarrow W$. Since $T(\mathbf{v}) = \mathbf{A}\mathbf{v}$ and

- $\{\mathbf{v}_1, \mathbf{v}_2\}$ is a basis for $\mathcal{C}(\mathbf{A}^T)$
- $\{\mathbf{v}_3, \mathbf{v}_4\}$ is a basis for $\mathcal{N}(\mathbf{A})$
- $\{\mathbf{w}_1, \mathbf{w}_2\}$ is a basis for $\mathcal{C}(\mathbf{A})$
- $\{\mathbf{w}_3\}$ is a basis for $\mathcal{N}(\mathbf{A}^T)$

we can have $\beta = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4\}$ and $\gamma = \{\mathbf{w}_1, \mathbf{w}_2, \mathbf{w}_3\}$ form a basis for V and W , respectively. As

$$\begin{aligned} T(\mathbf{v}_1) &= \mathbf{A}\mathbf{v}_1 = \mathbf{w}_1 \\ T(\mathbf{v}_2) &= \mathbf{A}\mathbf{v}_2 = \mathbf{w}_2 \\ T(\mathbf{v}_3) &= \mathbf{0} \\ T(\mathbf{v}_4) &= \mathbf{0} \end{aligned}$$

we can then obtain

$$[T]_{\beta}^{\gamma} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$

5. Based on the standard basis, the matrix which represents this T is

$$\begin{bmatrix} 0 & 2 & -1 \\ 2 & 3 & -2 \\ -1 & -2 & 0 \end{bmatrix}.$$

The eigenvectors for this matrix are $\{(-1, -2, 1), (-2, 1, 0), (1, 0, 1)\}$. Therefore, we can find the basis $\{(-1, -2, 1), (-2, 1, 0), (1, 0, 1)\}$ in which the matrix representation for T is a diagonal matrix.

6. (a) For all $\mathbf{x} \in \mathbb{R}^n$, we can have

$$\mathbf{x} = \mathbf{x}_r + \mathbf{x}_n$$

where $\mathbf{x}_r \in \mathcal{C}(\mathbf{A}^T)$ and $\mathbf{x}_n \in \mathcal{N}(\mathbf{A})$. We then have, for all $\mathbf{x} \in \mathbb{R}^n$,

$$\begin{aligned} \mathbf{A}\mathbf{A}^+\mathbf{A}\mathbf{x} &= \mathbf{A}\mathbf{A}^+\mathbf{A}(\mathbf{x}_r + \mathbf{x}_n) \\ &= \mathbf{A}\mathbf{A}^+\mathbf{A}\mathbf{x}_r + \mathbf{A}\mathbf{A}^+\mathbf{A}\mathbf{x}_n \\ &= \mathbf{A}(\mathbf{x}_r) + \mathbf{A}\mathbf{A}^+\mathbf{0} \\ &= \mathbf{A}\mathbf{x}_r \\ &= \mathbf{A}(\mathbf{x}_r + \mathbf{x}_n) \\ &= \mathbf{A}\mathbf{x} \end{aligned}$$

since $\mathbf{A}^+\mathbf{A}$ is the projection matrix onto $\mathcal{C}(\mathbf{A}^T)$ and $\mathbf{A}\mathbf{x}_n = \mathbf{0}$. Therefore, $\mathbf{A}\mathbf{A}^+\mathbf{A} = \mathbf{A}$

- (b) For a projection matrix \mathbf{P} , we have $\mathbf{P}^2 = \mathbf{P}$. Since $\mathbf{A}^+\mathbf{A}$ is the projection matrix onto $\mathcal{C}(\mathbf{A}^T)$,

$$(\mathbf{A}^+\mathbf{A})^2 = \mathbf{A}^+\mathbf{A}.$$

7. (a) After some calculations, we can obtain the eigenvalues and unit eigenvectors of $\mathbf{A}^T\mathbf{A}$ and $\mathbf{A}\mathbf{A}^T$ as follows:

$$\begin{aligned} \mathbf{A}^T\mathbf{A} = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 2 \\ 1 & 2 & 5 \end{bmatrix} &\implies \begin{cases} \lambda_1 = 6 & \longleftrightarrow & \mathbf{v}_1 = \frac{1}{\sqrt{30}}(1, 2, 5)^T \\ \lambda_2 = 1 & \longleftrightarrow & \mathbf{v}_2 = \frac{1}{\sqrt{5}}(-2, 1, 0)^T \\ \lambda_3 = 0 & \longleftrightarrow & \mathbf{v}_3 = \frac{1}{\sqrt{6}}(-1, -2, 1)^T. \end{cases} \\ \mathbf{A}\mathbf{A}^T = \begin{bmatrix} 5 & 2 \\ 2 & 2 \end{bmatrix} &\implies \begin{cases} \lambda_1 = 6 & \longleftrightarrow & \mathbf{u}_1 = \frac{1}{\sqrt{5}}(2, 1)^T \\ \lambda_2 = 1 & \longleftrightarrow & \mathbf{u}_2 = \frac{1}{\sqrt{5}}(1, -2)^T. \end{cases} \end{aligned}$$

The singular value decomposition of \mathbf{A} is hence given by

$$\mathbf{A} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$$

where

$$\mathbf{U} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & 1 \\ 1 & -2 \end{bmatrix}, \quad \mathbf{\Sigma} = \begin{bmatrix} \sqrt{6} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}, \quad \mathbf{V} = \frac{1}{\sqrt{30}} \begin{bmatrix} 1 & -2\sqrt{6} & -\sqrt{5} \\ 2 & \sqrt{6} & -2\sqrt{5} \\ 5 & 0 & \sqrt{5} \end{bmatrix}.$$

- (b) The pseudoinverse of \mathbf{A} is given by

$$\begin{aligned} \mathbf{A}^+ &= \mathbf{V}\mathbf{\Sigma}^+\mathbf{U}^T \\ &= \begin{bmatrix} -1/3 & 5/6 \\ 1/3 & -1/3 \\ 1/3 & 1/6 \end{bmatrix}. \end{aligned}$$

(c) The projection matrix onto the row space of \mathbf{A} is given by

$$\mathbf{A}^+ \mathbf{A} = \begin{bmatrix} 5/6 & -1/3 & 1/6 \\ -1/3 & 1/3 & 1/3 \\ 1/6 & 1/3 & 5/6 \end{bmatrix}.$$

(d) Since \mathbf{A} has full row rank, there is a right inverse for \mathbf{A} . We can have

$$\mathbf{A} \mathbf{A}^+ = \mathbf{I}$$

and hence the pseudoinverse \mathbf{A}^+ obtained in (b) is a right inverse for \mathbf{A} .

(e) The shortest least squares solution is

$$\mathbf{x} = \mathbf{A}^+ \mathbf{b} = \begin{bmatrix} -5/6 \\ 4/3 \\ 11/6 \end{bmatrix}.$$