

**Linear Algebra II, 2023 spring**  
**Midterm Exam Suggested Answers**

1. (a) For any  $\alpha, \beta \in \mathbb{R}$  and any  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^3$ ,

$$L(\alpha\mathbf{x} + \beta\mathbf{y}) = [2(\alpha x_2 + \beta y_2), -(\alpha x_1 + \beta y_1)] = \alpha [2x_2, -x_1] + \beta [2y_2, -y_1] = \alpha L(\mathbf{x}) + \beta L(\mathbf{y}).$$

Thus,  $L$  is a linear transformation from  $\mathbb{R}^3$  to  $\mathbb{R}^2$ .

- (b) Since

$$L(\mathbf{u}_1) = \begin{bmatrix} 0 \\ -1 \end{bmatrix} = -\frac{2}{3}\mathbf{b}_1 + \frac{1}{3}\mathbf{b}_2, \quad L(\mathbf{u}_2) = \begin{bmatrix} 4 \\ -1 \end{bmatrix} = \frac{2}{3}\mathbf{b}_1 + \frac{5}{3}\mathbf{b}_2$$

$$L(\mathbf{u}_3) = \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \frac{4}{3}\mathbf{b}_1 + \frac{1}{3}\mathbf{b}_2$$

Then

$$[L]_E^F = \frac{1}{3} \begin{bmatrix} -2 & 2 & 4 \\ 1 & 5 & 1 \end{bmatrix}.$$

2. (a) The rank of  $A$  is 2, then

$$\hat{\mathbf{x}} = (A^T A)^{-1} A^T \mathbf{b} = \begin{bmatrix} 2 \\ 1 \end{bmatrix},$$

where

$$A^T A = \begin{bmatrix} 6 & -3 \\ -3 & 6 \end{bmatrix}, \quad (A^T A)^{-1} = \frac{1}{9} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}, \quad A^T \mathbf{b} = \begin{bmatrix} 9 \\ 0 \end{bmatrix}.$$

- (b) Since  $\text{Col}(A)^\perp = \text{N}(A^T)$ , then

$$\text{rref}(A^T) = \text{rref} \begin{bmatrix} 1 & 2 & -1 & 0 \\ 0 & -1 & 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 & 4 \\ 0 & 1 & -1 & -2 \end{bmatrix} \rightarrow \text{N}(A^T) = \text{span} \left\{ \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -4 \\ 2 \\ 0 \\ 1 \end{bmatrix} \right\}$$

- (c) Using the Gram-Schmidt Process, let  $\mathbf{v}_1 = [-1 \ 1 \ 1 \ 0]^T$ , then

$$\mathbf{v}_2 = \begin{bmatrix} -4 \\ 2 \\ 0 \\ 1 \end{bmatrix} - \frac{(-4 \ 2 \ 0 \ 1) \begin{pmatrix} -1 \\ 1 \\ 1 \\ 0 \end{pmatrix}}{(-1 \ 1 \ 1 \ 0) \begin{pmatrix} -1 \\ 1 \\ 1 \\ 0 \end{pmatrix}} \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} -4 \\ 2 \\ 0 \\ 1 \end{bmatrix} - \frac{6}{3} \begin{bmatrix} -1 \\ 1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} -2 \\ 0 \\ -2 \\ 1 \end{bmatrix}.$$

Normalizing  $\mathbf{v}_1$  and  $\mathbf{v}_2$ , we have an orthonormal basis for  $\text{N}(A^T)$  as  $\left\{ \frac{1}{3} \begin{pmatrix} -1 \\ 1 \\ 1 \\ 0 \end{pmatrix}, \frac{1}{3} \begin{pmatrix} -4 \\ 2 \\ 0 \\ 1 \end{pmatrix} \right\}.$

3. The eigenvalues of  $A$  are  $\lambda_{1,2} = 1$  and  $\lambda_3 = 3$  since

$$\det(A - \lambda I_3) = (2 - \lambda)(1 - \lambda)(2 - \lambda) + (-1) \begin{vmatrix} 0 & 1 - \lambda \\ -1 & 0 \end{vmatrix} = (1 - \lambda)[(2 - \lambda)^2 - 1] = (1 - \lambda)^2(3 - \lambda) = 0.$$

For  $\lambda = 1$ ,  $(A - I_3)\mathbf{v}_1 = \mathbf{0}$ , i.e.  $\begin{pmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{pmatrix} \mathbf{v}_1 = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \mathbf{v}_1 = \alpha \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} + \beta \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$ ,  
 $\alpha, \beta$  are not all zero.

For  $\lambda = 3$ ,  $(A - 3I_3)\mathbf{v}_3 = \mathbf{0}$ , i.e.  $\begin{pmatrix} -1 & 0 & -1 \\ 0 & -1 & 0 \\ -1 & 0 & -1 \end{pmatrix} \mathbf{v}_3 = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \mathbf{v}_3 = \gamma \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \gamma \neq 0.$

4. Let  $\theta$  be the angle between  $Q\mathbf{x}$  and  $Q\mathbf{y}$ , and  $\alpha$  be the angle between  $\mathbf{x}$  and  $\mathbf{y}$ , then

$$\cos \theta = \frac{\langle Q\mathbf{x}, Q\mathbf{y} \rangle}{\|Q\mathbf{x}\| \|Q\mathbf{y}\|} = \frac{\mathbf{x}^T Q^T Q \mathbf{y}}{\|\mathbf{x}\| \|\mathbf{y}\|} = \frac{\mathbf{x}^T \mathbf{y}}{\|\mathbf{x}\| \|\mathbf{y}\|} = \cos \alpha,$$

since  $Q$  is orthogonal and  $Q^T Q = I$ , then  $\|Q\mathbf{z}\| = \sqrt{\mathbf{z}^T Q^T Q \mathbf{z}} = \sqrt{\mathbf{z}^T \mathbf{z}} = \|\mathbf{z}\|, \forall \mathbf{z} \in \mathbb{R}^n.$

5. (a)  $P^T = (A(A^T A)^{-1} A^T)^T = (A^T)^T (A^T A)^{-T} A^T = A(A^T A)^{-1} A^T = P.$   
(b) For every  $\mathbf{b} \in \text{Col}(A)$ ,  $\mathbf{b} = A\mathbf{y}$  for some  $\mathbf{y} \in \mathbb{R}^n$ ,

$$P\mathbf{b} = A(A^T A)^{-1} A^T \mathbf{b} = A(A^T A)^{-1} A^T A \mathbf{y} = A\mathbf{y} = \mathbf{b}.$$

Notice that  $P$  is the linear transformation of projecting a vector onto the column space of  $A$ , so any vector  $\mathbf{b}$  in  $\text{Col}(A)$ , it definitely will be projected to itself.

- (c) For any  $\mathbf{x} \in N(A^T)$ ,

$$P\mathbf{x} = A(A^T A)^{-1} A^T \mathbf{x} = A(A^T A)^{-1} [A^T \mathbf{x}] = A(A^T A)^{-1} \mathbf{0} = \mathbf{0}, \rightarrow \mathbf{x} \in N(P).$$

For any  $\mathbf{y} \in N(P)$ ,

$$A^T \mathbf{y} = A^T A(A^T A)^{-1} A^T \mathbf{y} = A^T P\mathbf{y} = \mathbf{0}, \rightarrow \mathbf{y} \in N(A^T).$$

Thus,  $N(A^T) = N(P).$

- (d) Based on the results in part (a) and (c),

$$\text{Col}(A) = N(A^T)^\perp = N(P)^\perp = \text{Col}(P^T) = \text{Col}(P)$$

6. Assume that there exists a nonsingular matrix  $S$  such that  $A = SBS^{-1}$ , then

$$\begin{aligned} \det(A - \lambda I) &= \det(SBS^{-1} - \lambda I) = \det(SBS^{-1} - \lambda S I S^{-1}) \\ &= \det[S(B - \lambda I)S^{-1}] = \det(S) \det(B - \lambda I) \det(S^{-1}) = \det(B - \lambda I) \end{aligned}$$

Thus,  $A$  and  $B$  have the same eigenvalues since they have the same characteristic polynomials.