

NATIONAL UNIVERSITY OF SINGAPORE
MATHEMATICS SOCIETY

PAST YEAR PAPER SOLUTIONS

Written by Lin Mingyan, Simon

Audited by Chua Hongshen

MA1101R Linear Algebra I
AY 2013/2014 Sem 1

Question 1

(i) We have

$$\mathbf{A}\mathbf{v} = \begin{pmatrix} 0 & 1 & 1 & 0 \\ -1 & 0 & 0 & -1 \\ 1 & 0 & 0 & 1 \\ 0 & -1 & -1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 2 \\ 2 \end{pmatrix} = \begin{pmatrix} 0 \cdot 1 + 1 \cdot 1 + 1 \cdot 2 + 0 \cdot 2 \\ (-1) \cdot 1 + 0 \cdot 1 + 0 \cdot 2 + (-1) \cdot 2 \\ 1 \cdot 1 + 0 \cdot 1 + 0 \cdot 2 + 1 \cdot 2 \\ 0 \cdot 1 + (-1) \cdot 1 + (-1) \cdot 2 + 0 \cdot 2 \end{pmatrix} = \begin{pmatrix} 3 \\ -3 \\ 3 \\ -3 \end{pmatrix} = \mathbf{b}_1.$$

(ii) By Gaussian elimination on the augmented matrix $(\mathbf{A}|\mathbf{0})$, we have

$$\begin{aligned} \left(\begin{array}{cccc|c} 0 & 1 & 1 & 0 & 0 \\ -1 & 0 & 0 & -1 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & -1 & -1 & 0 & 0 \end{array} \right) & \xrightarrow{R_2+R_3 \rightarrow R_2} \left(\begin{array}{cccc|c} 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & -1 & -1 & 0 & 0 \end{array} \right) \\ & \xrightarrow{R_1+R_4 \rightarrow R_4} \left(\begin{array}{cccc|c} 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \\ & \xrightarrow{R_1 \leftrightarrow R_2} \left(\begin{array}{cccc|c} 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \\ & \xrightarrow{R_1 \leftrightarrow R_3} \left(\begin{array}{cccc|c} 1 & 0 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right). \end{aligned}$$

So a basis for the nullspace of \mathbf{A} is $\{(-1, 0, 0, 1)^T, (0, -1, 1, 0)^T\}$.

(iii) The solution set for the matrix equation $\mathbf{A}\mathbf{x} = \mathbf{b}_1$ is $\{(1-s, 1-t, 2+t, 2+s)^T \mid s, t \in \mathbb{R}\}$.

(iv) By Gaussian elimination on the augmented matrix $(\mathbf{A}|\mathbf{b}_2)$, we have

$$\begin{aligned} \left(\begin{array}{cccc|c} 0 & 1 & 1 & 0 & 2 \\ -1 & 0 & 0 & -1 & 0 \\ 1 & 0 & 0 & 1 & -2 \\ 0 & -1 & -1 & 0 & 1 \end{array} \right) & \xrightarrow{R_2+R_3 \rightarrow R_2} \left(\begin{array}{cccc|c} 0 & 1 & 1 & 0 & 2 \\ 0 & 0 & 0 & 0 & -2 \\ 1 & 0 & 0 & 1 & -2 \\ 0 & -1 & -1 & 0 & 1 \end{array} \right) \\ & \xrightarrow{R_1+R_4 \rightarrow R_4} \left(\begin{array}{cccc|c} 0 & 1 & 1 & 0 & 2 \\ 0 & 0 & 0 & 0 & -2 \\ 1 & 0 & 0 & 1 & -2 \\ 0 & 0 & 0 & 0 & 3 \end{array} \right) \end{aligned}$$

$$\begin{aligned}
& \xrightarrow{R_2 + \frac{2}{3}R_4 \rightarrow R_2} \left(\begin{array}{cccc|c} 0 & 1 & 1 & 0 & 2 \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & -2 \\ 0 & 0 & 0 & 0 & 3 \end{array} \right) \\
& \xrightarrow{R_1 \leftrightarrow R_3} \left(\begin{array}{cccc|c} 1 & 0 & 0 & 1 & -2 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 2 \\ 0 & 0 & 0 & 0 & 3 \end{array} \right) \\
& \xrightarrow{R_2 \leftrightarrow R_3} \left(\begin{array}{cccc|c} 1 & 0 & 0 & 1 & -2 \\ 0 & 1 & 1 & 0 & 2 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 3 \end{array} \right) \\
& \xrightarrow{R_3 \leftrightarrow R_4} \left(\begin{array}{cccc|c} 1 & 0 & 0 & 1 & -2 \\ 0 & 1 & 1 & 0 & 2 \\ 0 & 0 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right).
\end{aligned}$$

Since $(\mathbf{A}|\mathbf{b}_2)$ has a row echelon form where the last column is a pivot column, we see that the matrix equation $\mathbf{A}\mathbf{x} = \mathbf{b}_2$ is inconsistent.

- (v) By part (ii), we have $B = \{(0, -1, 1, 0)^T, (1, 0, 0, -1)^T\}$ to be a basis for the column space of \mathbf{A} . Furthermore, it is clear that B is orthogonal. Let $\mathbf{u}_1 = (0, -1, 1, 0)^T$ and $\mathbf{u}_2 = (1, 0, 0, -1)^T$. Then the projection \mathbf{p} of \mathbf{b}_2 onto the column space of \mathbf{A} is given by

$$\begin{aligned}
\mathbf{p} &= \frac{\mathbf{u}_1 \cdot \mathbf{b}_2}{\|\mathbf{u}_1\|^2} \mathbf{u}_1 + \frac{\mathbf{u}_2 \cdot \mathbf{b}_2}{\|\mathbf{u}_2\|^2} \mathbf{u}_2 \\
&= \frac{(0, -1, 1, 0)^T \cdot (2, 0, -2, 1)^T}{2} (0, -1, 1, 0)^T + \frac{(1, 0, 0, -1)^T \cdot (2, 0, -2, 1)^T}{2} (1, 0, 0, -1)^T \\
&= \left(\frac{1}{2}, 1, -1, -\frac{1}{2} \right)^T.
\end{aligned}$$

By Gaussian elimination on the augmented matrix $(\mathbf{A}|\mathbf{p})$, we have

$$\begin{aligned}
& \left(\begin{array}{cccc|c} 0 & 1 & 1 & 0 & \frac{1}{2} \\ -1 & 0 & 0 & -1 & 1 \\ 1 & 0 & 0 & 1 & -1 \\ 0 & -1 & -1 & 0 & -\frac{1}{2} \end{array} \right) \xrightarrow{R_2 + R_3 \rightarrow R_2} \left(\begin{array}{cccc|c} 0 & 1 & 1 & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & -1 \\ 0 & -1 & -1 & 0 & -\frac{1}{2} \end{array} \right) \\
& \xrightarrow{R_1 + R_4 \rightarrow R_4} \left(\begin{array}{cccc|c} 0 & 1 & 1 & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \\
& \xrightarrow{R_1 \leftrightarrow R_3} \left(\begin{array}{cccc|c} 1 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \\
& \xrightarrow{R_2 \leftrightarrow R_3} \left(\begin{array}{cccc|c} 1 & 0 & 0 & 1 & -1 \\ 0 & 1 & 1 & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right).
\end{aligned}$$

Since a least square solution to the equation $\mathbf{A}\mathbf{x} = \mathbf{b}_2$ is a solution to the equation $\mathbf{A}\mathbf{x} = \mathbf{p}$, and vice versa, we see that two different least square solutions to the equation $\mathbf{A}\mathbf{x} = \mathbf{b}_2$ are

$\mathbf{x}_1 = (-1, \frac{1}{2}, 0, 0)^T$ and $\mathbf{x}_2 = (0, 0, \frac{1}{2}, -1)^T$. Lastly, we check that

$$\begin{aligned}
 \mathbf{Ax}_1 &= \begin{pmatrix} 0 & 1 & 1 & 0 \\ -1 & 0 & 0 & -1 \\ 1 & 0 & 0 & 1 \\ 0 & -1 & -1 & 0 \end{pmatrix} \begin{pmatrix} -1 \\ \frac{1}{2} \\ 0 \\ 0 \end{pmatrix} \\
 &= \begin{pmatrix} 0 \cdot (-1) + 1 \cdot \frac{1}{2} + 1 \cdot 0 + 0 \cdot 0 \\ (-1) \cdot (-1) + 0 \cdot \frac{1}{2} + 0 \cdot 0 + (-1) \cdot 0 \\ 1 \cdot (-1) + 0 \cdot \frac{1}{2} + 0 \cdot 0 + 1 \cdot 0 \\ 0 \cdot (-1) + (-1) \cdot \frac{1}{2} + (-1) \cdot 0 + 0 \cdot 0 \end{pmatrix} \\
 &= \begin{pmatrix} \frac{1}{2} \\ 1 \\ -1 \\ -\frac{1}{2} \end{pmatrix}, \text{ and} \\
 \mathbf{Ax}_2 &= \begin{pmatrix} 0 & 1 & 1 & 0 \\ -1 & 0 & 0 & -1 \\ 1 & 0 & 0 & 1 \\ 0 & -1 & -1 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ \frac{1}{2} \\ -1 \end{pmatrix} \\
 &= \begin{pmatrix} 0 \cdot 0 + 1 \cdot 0 + 1 \cdot \frac{1}{2} + 0 \cdot (-1) \\ (-1) \cdot 0 + 0 \cdot 0 + 0 \cdot \frac{1}{2} + (-1) \cdot (-1) \\ 1 \cdot 0 + 0 \cdot 0 + 0 \cdot \frac{1}{2} + 1 \cdot (-1) \\ 0 \cdot 0 + (-1) \cdot 0 + (-1) \cdot \frac{1}{2} + 0 \cdot (-1) \end{pmatrix} \\
 &= \begin{pmatrix} \frac{1}{2} \\ 1 \\ -1 \\ -\frac{1}{2} \end{pmatrix},
 \end{aligned}$$

so we have $\mathbf{Ax}_1 = \mathbf{Ax}_2$ as desired.

- (vi) It is not possible for the matrix equation $\mathbf{Ax} = \mathbf{b}_1 + \mathbf{b}_3$ to be consistent. Note that for a given $\mathbf{a} \in \mathbb{R}^4$, we have the equation $\mathbf{Ax} = \mathbf{a}$ to be consistent if and only if \mathbf{a} belongs to the column space of \mathbf{A} . Arguing by contradiction, suppose that the equation $\mathbf{Ax} = \mathbf{b}_1 + \mathbf{b}_3$ is consistent. Then $\mathbf{b}_1 + \mathbf{b}_3$ must belong to the column space of \mathbf{A} . As the column space of \mathbf{A} is a vector subspace of \mathbb{R}^4 , and \mathbf{b}_1 belongs to the column space of \mathbf{A} by part (a), we must have $\mathbf{b}_3 = (\mathbf{b}_1 + \mathbf{b}_3) - \mathbf{b}_1$ to belong to the column space of \mathbf{A} , a contradiction.

Question 2

By Gaussian elimination on the augmented matrix $(\mathbf{X}|\mathbf{I})$, we have

$$\begin{aligned}
 \left(\begin{array}{ccc|ccc} 2 & 0 & 0 & 1 & 0 & 0 \\ 1 & 2 & -1 & 0 & 1 & 0 \\ 1 & 3 & -2 & 0 & 0 & 1 \end{array} \right) &\xrightarrow{\frac{1}{2}R_2 \rightarrow R_2} \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{1}{2} & 0 & 0 \\ 1 & 2 & -1 & 0 & 1 & 0 \\ 1 & 3 & -2 & 0 & 0 & 1 \end{array} \right) \\
 &\xrightarrow{R_2 - R_1 \rightarrow R_2} \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{1}{2} & 0 & 0 \\ 0 & 2 & -1 & -\frac{1}{2} & 1 & 0 \\ 1 & 3 & -2 & 0 & 0 & 1 \end{array} \right) \\
 &\xrightarrow{R_3 - R_1 \rightarrow R_3} \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{1}{2} & 0 & 0 \\ 0 & 2 & -1 & -\frac{1}{2} & 1 & 0 \\ 0 & 3 & -2 & -\frac{1}{2} & 0 & 1 \end{array} \right) \\
 &\xrightarrow{R_2 - \frac{1}{2}R_3 \rightarrow R_2} \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{1}{2} & 0 & 0 \\ 0 & \frac{1}{2} & 0 & -\frac{1}{4} & 1 & -\frac{1}{2} \\ 0 & 3 & -2 & -\frac{1}{2} & 0 & 1 \end{array} \right)
 \end{aligned}$$

$$\begin{aligned}
&\xrightarrow{2R_2 \rightarrow R_2} \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 & -\frac{1}{2} & 2 & -1 \\ 0 & 3 & -2 & -\frac{1}{2} & 0 & 1 \end{array} \right) \\
&\xrightarrow{R_3 - 3R_2 \rightarrow R_3} \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 & -\frac{1}{2} & 2 & -1 \\ 0 & 0 & -2 & 1 & -6 & 4 \end{array} \right) \\
&\xrightarrow{-\frac{1}{2}R_3 \rightarrow R_3} \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 & -\frac{1}{2} & 2 & -1 \\ 0 & 0 & 1 & -\frac{1}{2} & 3 & -2 \end{array} \right).
\end{aligned}$$

Hence, we have $\mathbf{X}^{-1} = \begin{pmatrix} \frac{1}{2} & 0 & 0 \\ -\frac{1}{2} & 2 & -1 \\ -\frac{1}{2} & 3 & -2 \end{pmatrix}$.

- (ii) We have $\det(\mathbf{X}) = 2(2)(-2) + 0(-1)(1) + 0(1)(3) - 0(2)(1) - 0(1)(-2) - 2(-1)(3) = -2$. Thus, we have

$$\mathbf{adj}(\mathbf{X}) = \det(\mathbf{X})\mathbf{X}^{-1} = -2 \begin{pmatrix} \frac{1}{2} & 0 & 0 \\ -\frac{1}{2} & 2 & -1 \\ -\frac{1}{2} & 3 & -2 \end{pmatrix} = \begin{pmatrix} -1 & 0 & 0 \\ 1 & -4 & 2 \\ 1 & -6 & 4 \end{pmatrix}.$$

- (iii) Let us denote the standard matrix for the linear transformation $S : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ by \mathbf{Z} . Since we have $(T \circ S)(\mathbf{x}) = \mathbf{x}$ for all non-zero $\mathbf{x} \in \mathbb{R}^3$, it follows that $T \circ S = I$, where I denotes the identity transformation on \mathbb{R}^3 . Hence, the standard matrix for the linear transformation $T \circ S : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ is the identity matrix \mathbf{I} , which implies that $\mathbf{Z}\mathbf{X} = \mathbf{I}$. Hence, we have $\mathbf{Z} = \mathbf{X}^{-1}$, so the formula for the linear transformation $S : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ is given by

$$S \left(\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \right) = \begin{pmatrix} \frac{1}{2} & 0 & 0 \\ -\frac{1}{2} & 2 & -1 \\ -\frac{1}{2} & 3 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} \frac{x_1}{2} \\ -\frac{x_1}{2} + 2x_2 - x_3 \\ -\frac{x_1}{2} + 3x_2 - 2x_3 \end{pmatrix}$$

for all $(x_1, x_2, x_3)^T \in \mathbb{R}^3$.

- (iv) Note that

$$\lambda \mathbf{I} - \mathbf{X} = \begin{pmatrix} \lambda - 2 & 0 & 0 \\ -1 & \lambda - 2 & 1 \\ -1 & -3 & \lambda + 2 \end{pmatrix}$$

for all $\lambda \in \mathbb{R}$. This implies that

$$\begin{aligned}
\det(\lambda \mathbf{I} - \mathbf{X}) &= (\lambda - 2)(\lambda - 2)(\lambda + 2) - (\lambda - 2)(1)(-3) \\
&= (\lambda - 2)^2(\lambda + 2) + 3(\lambda - 2) \\
&= (\lambda - 2)((\lambda - 2)(\lambda + 2) + 3) \\
&= (\lambda - 2)(\lambda^2 - 4 + 3) \\
&= (\lambda - 2)(\lambda^2 - 1) \\
&= (\lambda - 2)(\lambda - 1)(\lambda + 1).
\end{aligned}$$

Note that $\det(\lambda \mathbf{I} - \mathbf{X}) = 0$ if and only if $\lambda = 2, 1, -1$. So the eigenvalues λ of \mathbf{X} are $\lambda = 2, 1, -1$.

When $\lambda = 2$, we see from Gaussian elimination on the augmented matrix $(\lambda \mathbf{I} - \mathbf{X} | \mathbf{0})$ that

$$\left(\begin{array}{ccc|c} 0 & 0 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & -3 & 4 & 0 \end{array} \right) \xrightarrow{-R_2 + R_3 \rightarrow R_3} \left(\begin{array}{ccc|c} 0 & 0 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ 0 & -3 & 3 & 0 \end{array} \right)$$

$$\begin{aligned} \xrightarrow{R_1 \leftrightarrow R_2} & \left(\begin{array}{ccc|c} -1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & -3 & 3 & 0 \end{array} \right) \\ \xrightarrow{R_2 \leftrightarrow R_3} & \left(\begin{array}{ccc|c} -1 & 0 & 1 & 0 \\ 0 & -3 & 3 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right). \end{aligned}$$

So a basis for E_2 is $\{(1, 1, 1)^T\}$.

When $\lambda = 1$, we see from Gaussian elimination on the augmented matrix $(\lambda \mathbf{I} - \mathbf{X} | \mathbf{0})$ that

$$\begin{aligned} \left(\begin{array}{ccc|c} -1 & 0 & 0 & 0 \\ -1 & -1 & 1 & 0 \\ -1 & -3 & 3 & 0 \end{array} \right) & \xrightarrow{-R_1 + R_2 \rightarrow R_2} \left(\begin{array}{ccc|c} -1 & 0 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ -1 & -3 & 3 & 0 \end{array} \right) \\ & \xrightarrow{-R_1 + R_3 \rightarrow R_3} \left(\begin{array}{ccc|c} -1 & 0 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & -3 & 3 & 0 \end{array} \right) \\ & \xrightarrow{-3R_2 + R_3 \rightarrow R_3} \left(\begin{array}{ccc|c} -1 & 0 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right). \end{aligned}$$

So a basis for E_1 is $\{(0, 1, 1)^T\}$.

When $\lambda = -1$, we see from Gaussian elimination on the augmented matrix $(\lambda \mathbf{I} - \mathbf{X} | \mathbf{0})$ that

$$\begin{aligned} \left(\begin{array}{ccc|c} -3 & 0 & 0 & 0 \\ -1 & -3 & 1 & 0 \\ -1 & -3 & 1 & 0 \end{array} \right) & \xrightarrow{-R_2 + R_3 \rightarrow R_3} \left(\begin{array}{ccc|c} -3 & 0 & 0 & 0 \\ -1 & -3 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right) \\ & \xrightarrow{-\frac{1}{3}R_1 + R_2 \rightarrow R_2} \left(\begin{array}{ccc|c} -3 & 0 & 0 & 0 \\ 0 & -3 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right). \end{aligned}$$

So a basis for E_{-1} is $\{(0, 1, 3)^T\}$.

- (v) Since \mathbf{X} has 3 distinct eigenvalues, \mathbf{X} is diagonalizable, and a matrix \mathbf{P} that diagonalizes \mathbf{X} is

$$\mathbf{P} = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & 1 & 3 \end{pmatrix}.$$

- (vi) Clearly, we have $E_{\lambda_1} + E_{\lambda_2} + \cdots + E_{\lambda_k} \subseteq \mathbb{R}^n$, so it remains to show that $\mathbb{R}^n \subseteq E_{\lambda_1} + E_{\lambda_2} + \cdots + E_{\lambda_k}$. To this end, write $S_{\lambda_i} = \{\mathbf{u}_{i,1}, \mathbf{u}_{i,2}, \dots, \mathbf{u}_{i,n_i}\}$ for all $i = 1, 2, \dots, k$, where n_i is a positive integer for all $i = 1, 2, \dots, k$. Furthermore, let $S = \bigcup_{i=1}^k S_{\lambda_i}$. Then S is linearly independent, since the S_{λ_i} 's are chosen to be bases for the E_{λ_i} 's for all $i = 1, 2, \dots, k$. As \mathbf{Y} is a diagonalizable matrix of order n , we must have $|S| = n$. Hence, we have S to be a basis for \mathbb{R}^n .

Now, let us take any $\mathbf{x} \in \mathbb{R}^n$. As S is a basis for \mathbb{R}^n , it follows that for all $i = 1, 2, \dots, k$ and $j = 1, 2, \dots, n_i$, there exist (unique) $a_{i,j} \in \mathbb{R}$, such that $\mathbf{x} = \sum_{i=1}^k \left(\sum_{j=1}^{n_i} a_{i,j} \mathbf{u}_{i,j} \right)$. As E_{λ_i} is a vector subspace of \mathbb{R}^n for all $i = 1, 2, \dots, k$, we must have $\mathbf{v}_i := \sum_{j=1}^{n_i} a_{i,j} \mathbf{u}_{i,j} \in E_{\lambda_i}$ for all $i = 1, 2, \dots, k$.

This implies that $\mathbf{x} = \sum_{i=1}^k \left(\sum_{j=1}^{n_i} a_{i,j} \mathbf{u}_{i,j} \right) = \sum_{i=1}^k \mathbf{v}_i \in E_{\lambda_1} + E_{\lambda_2} + \cdots + E_{\lambda_k}$. As $\mathbf{x} \in \mathbb{R}^n$ is arbitrary, this shows that $\mathbb{R}^n \subseteq E_{\lambda_1} + E_{\lambda_2} + \cdots + E_{\lambda_k}$, and we are done.

Question 3

- (i) Let \mathbf{A} denote the matrix $(\mathbf{u}_1 \mathbf{u}_2 \mathbf{u}_3)^T$. By Gaussian elimination on the augmented matrix $(\mathbf{A}|\mathbf{0})$, we have

$$\left(\begin{array}{cccc|c} 1 & 1 & 0 & 2 & 0 \\ 0 & -2 & 1 & 1 & 0 \\ 0 & 1 & 1 & 2 & 0 \end{array}\right) \xrightarrow{\frac{1}{2}R_2+R_3 \rightarrow R_3} \left(\begin{array}{cccc|c} 1 & 1 & 0 & 2 & 0 \\ 0 & -2 & 1 & 1 & 0 \\ 0 & 0 & \frac{3}{2} & \frac{5}{2} & 0 \end{array}\right).$$

Since the augmented matrix $(\mathbf{A}|\mathbf{0})$ has a row echelon form where all rows are non-zero, we see that S is linearly independent, and hence S is a basis for V .

- (ii) Let \mathbf{B} denote the matrix $(\mathbf{u}_1^T \mathbf{u}_2^T \mathbf{u}_3^T)$. By Gaussian elimination on the augmented matrix $(\mathbf{B}|\mathbf{w}^T)$, we have

$$\begin{aligned} \left(\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 1 & -2 & 1 & -1 \\ 0 & 1 & 1 & -1 \\ 2 & 1 & 2 & 3 \end{array}\right) & \xrightarrow{-R_1+R_2 \rightarrow R_2} \left(\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & -2 & 1 & -4 \\ 0 & 1 & 1 & -1 \\ 2 & 1 & 2 & 3 \end{array}\right) \\ & \xrightarrow{-2R_1+R_4 \rightarrow R_4} \left(\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & -2 & 1 & -4 \\ 0 & 1 & 1 & -1 \\ 0 & 1 & 2 & -3 \end{array}\right) \\ & \xrightarrow{2R_3+R_2 \rightarrow R_2} \left(\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & 0 & 3 & -6 \\ 0 & 1 & 1 & -1 \\ 0 & 1 & 2 & -3 \end{array}\right) \\ & \xrightarrow{-R_3+R_4 \rightarrow R_4} \left(\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & 0 & 3 & -6 \\ 0 & 1 & 1 & -1 \\ 0 & 0 & 1 & -2 \end{array}\right) \\ & \xrightarrow{-3R_4+R_2 \rightarrow R_2} \left(\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & -1 \\ 0 & 0 & 1 & -2 \end{array}\right) \\ & \xrightarrow{-R_4+R_3 \rightarrow R_3} \left(\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & -2 \end{array}\right) \\ & \xrightarrow{R_2 \leftrightarrow R_3} \left(\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & -2 \end{array}\right) \\ & \xrightarrow{R_3 \leftrightarrow R_4} \left(\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & -2 \\ 0 & 0 & 0 & 0 \end{array}\right). \end{aligned}$$

Since the last row of the last augmented matrix in the above equation is zero, it follows that $\mathbf{w} \in V$, and $(\mathbf{w})_S = (3, 1, -2)$.

- (iii) By Gram-Schmidt process on $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$, we have

$$\mathbf{v}_1 := \mathbf{u}_1$$

$$\begin{aligned}
&= (1, 1, 0, 2), \\
\mathbf{v}_2 &:= \mathbf{u}_2 - \frac{\mathbf{u}_2 \cdot \mathbf{v}_1}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 \\
&= (0, -2, 1, 1) - \frac{(0, -2, 1, 1) \cdot (1, 1, 0, 2)}{1^2 + 1^2 + 2^2} (1, 1, 0, 2) \\
&= (0, -2, 1, 1), \\
\mathbf{v}_3 &:= \mathbf{u}_3 - \frac{\mathbf{u}_3 \cdot \mathbf{v}_1}{\|\mathbf{v}_1\|^2} \mathbf{v}_1 - \frac{\mathbf{u}_3 \cdot \mathbf{v}_2}{\|\mathbf{v}_2\|^2} \mathbf{v}_2 \\
&= (0, 1, 1, 2) - \frac{(0, 1, 1, 2) \cdot (1, 1, 0, 2)}{1^2 + 1^2 + 2^2} (1, 1, 0, 2) - \frac{(0, 1, 1, 2) \cdot (0, -2, 1, 1)}{(-2)^2 + 1^2 + 1^2} (0, -2, 1, 1) \\
&= \left(-\frac{5}{6}, \frac{1}{2}, \frac{5}{6}, \frac{1}{6} \right).
\end{aligned}$$

So an orthogonal basis T for V is $\{(1, 1, 0, 2), (0, -2, 1, 1), (-\frac{5}{6}, \frac{1}{2}, \frac{5}{6}, \frac{1}{6})\}$.

- (iv) By assumption, we have $(0, 1, 1, 2) = \frac{5}{6}(1, 1, 0, 2) + \frac{1}{6}(0, -2, 1, 1) + (-\frac{5}{6}, \frac{1}{2}, \frac{5}{6}, \frac{1}{6})$, so this implies that

$$\begin{aligned}
&(3, -1, -1, 3) \\
&= 3(1, 1, 0, 2) + (0, -2, 1, 1) - 2(0, 1, 1, 2) \\
&= 3(1, 1, 0, 2) + (0, -2, 1, 1) - 2\left(\frac{5}{6}(1, 1, 0, 2) + \frac{1}{6}(0, -2, 1, 1) + \left(-\frac{5}{6}, \frac{1}{2}, \frac{5}{6}, \frac{1}{6}\right)\right) \\
&= \frac{4}{3}(1, 1, 0, 2) + \frac{2}{3}(0, -2, 1, 1) - 2\left(-\frac{5}{6}, \frac{1}{2}, \frac{5}{6}, \frac{1}{6}\right).
\end{aligned}$$

Hence, we have $(\mathbf{w})_T = (\frac{4}{3}, \frac{2}{3}, -2)$.

- (v) Note that a row echelon form for the matrix \mathbf{A} is $\begin{pmatrix} 1 & 1 & 0 & 2 \\ 0 & -2 & 1 & 1 \\ 0 & 0 & \frac{3}{2} & \frac{5}{2} \end{pmatrix}$. Since the last column

in the above row echelon form of \mathbf{A} is a non-pivot column, we see that a basis for \mathbb{R}^4 that contains S is $\{(1, 1, 0, 2), (0, -2, 1, 1), (0, 1, 1, 2), (0, 0, 0, 1)\}$.

- (vi) Remark. We note that if the kernel of a linear transformation $T : \mathbb{R}^4 \rightarrow \mathbb{R}^2$ is V , then by the Rank-Nullity Theorem, we must have $\text{rank}(T) + \text{nullity}(T) = 4$, which implies that $\text{rank}(T) = 1$, since $\text{nullity}(T) = \dim(\text{Ker}(T)) = \dim(V) = 3$. As

$$\{(1, 1, 0, 2), (0, -2, 1, 1), (0, 1, 1, 2), (0, 0, 0, 1)\}$$

is a basis for \mathbb{R}^4 that contains S , and S is a basis for V , the strategy to constructing a linear transformation $T : \mathbb{R}^4 \rightarrow \mathbb{R}^2$ that satisfies the given conditions is to first set the images of each of the elements in S under T to $\mathbf{0}$, and set the image of $(0, 0, 0, 1)$ under T to any non-zero element of \mathbb{R}^2 . The formula for T could be found thereafter using Gaussian elimination.

Let $\mathbf{u}_4 = \mathbf{e}_4 = (0, 0, 0, 1)$, and \mathbf{C} denote the matrix $(\mathbf{u}_1^T \mathbf{u}_2^T \mathbf{u}_3^T \mathbf{u}_4^T)$. By Gaussian elimination on the augmented matrix $(\mathbf{C} | \mathbf{e}_1 | \mathbf{e}_2 | \mathbf{e}_3)$, we have

$$\begin{aligned}
\left(\begin{array}{cccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 1 & -2 & 1 & 0 & 0 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 0 & 1 \\ 2 & 1 & 2 & 1 & 0 & 0 & 0 \end{array} \right) &\xrightarrow{-R_1+R_2 \rightarrow R_2} \left(\begin{array}{cccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & -2 & 1 & 0 & -1 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 0 & 1 \\ 2 & 1 & 2 & 1 & 0 & 0 & 0 \end{array} \right) \\
&\xrightarrow{-2R_1+R_4 \rightarrow R_4} \left(\begin{array}{cccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & -2 & 1 & 0 & -1 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 0 & 1 \\ 0 & 1 & 2 & 1 & -2 & 0 & 0 \end{array} \right)
\end{aligned}$$

$$\begin{aligned}
&\xrightarrow{R_2 - R_3 \rightarrow R_2} \left(\begin{array}{cccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & -3 & 0 & 0 & -1 & 1 & -1 \\ 0 & 1 & 1 & 0 & 0 & 0 & 1 \\ 0 & 1 & 2 & 1 & -2 & 0 & 0 \end{array} \right) \\
&\xrightarrow{-\frac{1}{3}R_2 \rightarrow R_2} \left(\begin{array}{cccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & \frac{1}{3} & -\frac{1}{3} & \frac{1}{3} \\ 0 & 1 & 1 & 0 & 0 & 0 & 1 \\ 0 & 1 & 2 & 1 & -2 & 0 & 0 \end{array} \right) \\
&\xrightarrow{-R_2 + R_3 \rightarrow R_3} \left(\begin{array}{cccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & \frac{1}{3} & -\frac{1}{3} & \frac{1}{3} \\ 0 & 0 & 1 & 0 & -\frac{1}{3} & \frac{1}{3} & \frac{2}{3} \\ 0 & 1 & 2 & 1 & -2 & 0 & 0 \end{array} \right) \\
&\xrightarrow{-R_2 + R_4 \rightarrow R_4} \left(\begin{array}{cccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & \frac{1}{3} & -\frac{1}{3} & \frac{1}{3} \\ 0 & 0 & 1 & 0 & -\frac{1}{3} & \frac{1}{3} & \frac{2}{3} \\ 0 & 0 & 2 & 1 & -\frac{5}{3} & \frac{1}{3} & -\frac{1}{3} \end{array} \right) \\
&\xrightarrow{-2R_3 + R_4 \rightarrow R_4} \left(\begin{array}{cccc|ccc} 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & \frac{1}{3} & -\frac{1}{3} & \frac{1}{3} \\ 0 & 0 & 1 & 0 & -\frac{1}{3} & \frac{1}{3} & \frac{2}{3} \\ 0 & 0 & 0 & 1 & -\frac{5}{3} & \frac{1}{3} & -\frac{5}{3} \end{array} \right).
\end{aligned}$$

Now, let us define a linear transformation $T : \mathbb{R}^4 \rightarrow \mathbb{R}^2$ by

$$T(x_1, x_2, x_3, x_4) = \left(-\frac{5x_1}{3} - \frac{x_2}{3} - \frac{5x_3}{3} + x_4, 0 \right)$$

for all $(x_1, x_2, x_3, x_4) \in \mathbb{R}^4$. Let us show that $\text{Ker}(T) = V$. Since

$$\begin{aligned}
T(1, 1, 0, 2) &= \left(-\frac{5}{3} - \frac{1}{3} + 0 + 2, 0 \right) = (0, 0), \\
T(0, -2, 1, 1) &= \left(0 + \frac{2}{3} - \frac{5}{3} + 1, 0 \right) = (0, 0), \text{ and} \\
T(0, 1, 1, 2) &= \left(0 - \frac{1}{3} - \frac{5}{3} + 2, 0 \right) = (0, 0),
\end{aligned}$$

we see that $S \subseteq \text{Ker}(T)$. As $\text{Ker}(T)$ is a vector subspace of \mathbb{R}^4 , we must have $V = \text{span}(S) \subseteq \text{Ker}(T)$, so that $\text{nullity}(T) = \dim(\text{Ker}(T)) \geq \dim(V) = 3$. Next, since $(0, 0, 0, 1) \notin \text{Ker}(T)$, we must have $\text{Ker}(T) \neq \mathbb{R}^4$, so we must have $\text{nullity}(T) < 4$. So $\text{nullity}(T) = 3$, and hence we must have $\text{Ker}(T) = V$ as desired.

Question 4

(i) By Gaussian elimination on \mathbf{Y} , we have

$$\begin{aligned}
&\left(\begin{array}{cccc} 2 & -4 & 6 & 8 \\ 2 & -1 & 3 & 2 \\ 4 & -5 & 9 & 10 \\ 0 & -1 & 1 & 2 \end{array} \right) \xrightarrow{-R_1 + R_2 \rightarrow R_2} \left(\begin{array}{cccc} 2 & -4 & 6 & 8 \\ 0 & 3 & -3 & -6 \\ 4 & -5 & 9 & 10 \\ 0 & -1 & 1 & 2 \end{array} \right) \\
&\xrightarrow{-2R_1 + R_3 \rightarrow R_3} \left(\begin{array}{cccc} 2 & -4 & 6 & 8 \\ 0 & 3 & -3 & -6 \\ 0 & 3 & -3 & -6 \\ 0 & -1 & 1 & 2 \end{array} \right)
\end{aligned}$$

$$\begin{aligned}
& \xrightarrow{3R_4+R_2 \rightarrow R_2} \begin{pmatrix} 2 & -4 & 6 & 8 \\ 0 & 0 & 0 & 0 \\ 0 & 3 & -3 & -6 \\ 0 & -1 & 1 & 2 \end{pmatrix} \\
& \xrightarrow{3R_4+R_3 \rightarrow R_3} \begin{pmatrix} 2 & -4 & 6 & 8 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & -1 & 1 & 2 \end{pmatrix} \\
& \xrightarrow{R_2 \leftrightarrow R_4} \begin{pmatrix} 2 & -4 & 6 & 8 \\ 0 & -1 & 1 & 2 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}.
\end{aligned}$$

Hence, we have $\text{rank}(\mathbf{Y}) = 2$ as required.

(ii) By Rank-Nullity Theorem, we have $\text{nullity}(\mathbf{Y}) = 4 - \text{rank}(\mathbf{Y}) = 2$. Moreover, we have

$$\begin{aligned}
\mathbf{Y}\mathbf{v}_1 &= \begin{pmatrix} 2 & -4 & 6 & 8 \\ 2 & -1 & 3 & 2 \\ 4 & -5 & 9 & 10 \\ 0 & -1 & 1 & 2 \end{pmatrix} \begin{pmatrix} 2 \\ 0 \\ -2 \\ 1 \end{pmatrix} = \begin{pmatrix} 2 \cdot 2 + (-4) \cdot 0 + 6 \cdot (-2) + 8 \cdot 1 \\ 2 \cdot 2 + (-1) \cdot 0 + 3 \cdot (-2) + 2 \cdot 1 \\ 4 \cdot 2 + (-5) \cdot 0 + 9 \cdot (-2) + 10 \cdot 1 \\ 0 \cdot 2 + (-1) \cdot 0 + 1 \cdot (-2) + 2 \cdot 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \text{ and} \\
\mathbf{Y}\mathbf{v}_2 &= \begin{pmatrix} 2 & -4 & 6 & 8 \\ 2 & -1 & 3 & 2 \\ 4 & -5 & 9 & 10 \\ 0 & -1 & 1 & 2 \end{pmatrix} \begin{pmatrix} 1 \\ 5 \\ -1 \\ 3 \end{pmatrix} = \begin{pmatrix} 2 \cdot 1 + (-4) \cdot 5 + 6 \cdot (-1) + 8 \cdot 3 \\ 2 \cdot 1 + (-1) \cdot 5 + 3 \cdot (-1) + 2 \cdot 3 \\ 4 \cdot 1 + (-5) \cdot 5 + 9 \cdot (-1) + 10 \cdot 3 \\ 0 \cdot 1 + (-1) \cdot 5 + 1 \cdot (-1) + 2 \cdot 3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}.
\end{aligned}$$

This implies that $\{\mathbf{v}_1, \mathbf{v}_2\}$ is contained in the nullspace of \mathbf{Y} . Finally, since $\{\mathbf{v}_1, \mathbf{v}_2\}$ is clearly linearly independent, and $\text{nullity}(\mathbf{Y}) = 2$, we must have $\{\mathbf{v}_1, \mathbf{v}_2\}$ to be a basis for the nullspace of \mathbf{Y} as desired.

(iii) By Gaussian elimination on the augmented matrix $(\mathbf{Z}|\mathbf{v}_1)$, we have

$$\begin{aligned}
& \left(\begin{array}{cccc|c} 1 & 1 & 0 & 0 & 2 \\ 0 & 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 0 & -2 \\ 0 & 1 & 0 & 1 & 1 \end{array} \right) \xrightarrow{-R_1+R_3 \rightarrow R_3} \left(\begin{array}{cccc|c} 1 & 1 & 0 & 0 & 2 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & -1 & 1 & 0 & -4 \\ 0 & 1 & 0 & 1 & 1 \end{array} \right) \\
& \xrightarrow{R_3+R_4 \rightarrow R_4} \left(\begin{array}{cccc|c} 1 & 1 & 0 & 0 & 2 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & -1 & 1 & 0 & -4 \\ 0 & 0 & 1 & 1 & -3 \end{array} \right) \\
& \xrightarrow{-R_2+R_4 \rightarrow R_4} \left(\begin{array}{cccc|c} 1 & 1 & 0 & 0 & 2 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & -1 & 1 & 0 & -4 \\ 0 & 0 & 0 & 0 & -3 \end{array} \right) \\
& \xrightarrow{R_2 \leftrightarrow R_4} \left(\begin{array}{cccc|c} 1 & 1 & 0 & 0 & 2 \\ 0 & -1 & 1 & 0 & -4 \\ 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & -3 \end{array} \right).
\end{aligned}$$

Since $(\mathbf{Z}|\mathbf{v}_1)$ has a row echelon form where the last column is a pivot column, we see that the matrix equation $\mathbf{Z}\mathbf{x} = \mathbf{v}_1$ is inconsistent.

Now, we note that $B = \{(1, 0, 1, 0)^T, (1, 0, 0, 1)^T, (0, 1, 1, 0)^T\}$ to be a basis for the column space of \mathbf{Z} . Let $\mathbf{u}_1 = (0, 1, 1, 0)^T$, $\mathbf{u}_2 = (1, 0, 0, 1)^T$ and $\mathbf{u}_3 = (1, 0, 1, 0)^T$. Then it is clear

that \mathbf{u}_1 and \mathbf{u}_2 are orthogonal to each other. By Gram-Schmidt process on $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$, we have

$$\begin{aligned} & \mathbf{u}_3 - \frac{\mathbf{u}_3 \cdot \mathbf{u}_1}{\|\mathbf{u}_1\|^2} \mathbf{u}_1 - \frac{\mathbf{u}_3 \cdot \mathbf{u}_2}{\|\mathbf{u}_2\|^2} \mathbf{u}_2 \\ = & (1, 0, 1, 0)^T - \frac{(1, 0, 1, 0)^T \cdot (0, 1, 1, 0)^T}{1^2 + 1^2} (0, 1, 1, 0)^T - \frac{(1, 0, 1, 0)^T \cdot (1, 0, 0, 1)^T}{1^2 + 1^2} (1, 0, 0, 1)^T \\ = & \left(\frac{1}{2}, -\frac{1}{2}, \frac{1}{2}, -\frac{1}{2} \right)^T. \end{aligned}$$

So an orthogonal basis B' for the column space of \mathbf{Z} is

$$\left\{ (0, 1, 1, 0)^T, (1, 0, 0, 1)^T, \left(\frac{1}{2}, -\frac{1}{2}, \frac{1}{2}, -\frac{1}{2} \right)^T \right\}.$$

Now, let \mathbf{p} be the projection of \mathbf{v}_1 onto the column space of \mathbf{Z} , and $\mathbf{w}_1 = (0, 1, 1, 0)^T$, $\mathbf{w}_2 = (1, 0, 0, 1)^T$ and $\mathbf{w}_3 = \left(\frac{1}{2}, -\frac{1}{2}, \frac{1}{2}, -\frac{1}{2} \right)^T$. Now, we have

$$\begin{aligned} \mathbf{w}_1 \cdot \mathbf{v}_1 &= (0, 1, 1, 0)^T \cdot (2, 0, -2, 1)^T = -2, \\ \mathbf{w}_2 \cdot \mathbf{v}_1 &= (1, 0, 0, 1)^T \cdot (2, 0, -2, 1)^T = 3, \\ \mathbf{w}_3 \cdot \mathbf{v}_1 &= \left(\frac{1}{2}, -\frac{1}{2}, \frac{1}{2}, -\frac{1}{2} \right)^T \cdot (2, 0, -2, 1)^T = -\frac{1}{2}, \\ \mathbf{p} &= \frac{\mathbf{w}_1 \cdot \mathbf{v}_1}{\|\mathbf{w}_1\|^2} \mathbf{w}_1 + \frac{\mathbf{w}_2 \cdot \mathbf{v}_1}{\|\mathbf{w}_2\|^2} \mathbf{w}_2 + \frac{\mathbf{w}_3 \cdot \mathbf{v}_1}{\|\mathbf{w}_3\|^2} \mathbf{w}_3 \\ &= \frac{-2}{2} (0, 1, 1, 0)^T + \frac{3}{2} (1, 0, 0, 1)^T - \frac{1}{2} \left(\frac{1}{2}, -\frac{1}{2}, \frac{1}{2}, -\frac{1}{2} \right)^T \\ &= \left(\frac{5}{4}, -\frac{3}{4}, -\frac{5}{4}, \frac{7}{4} \right)^T. \end{aligned}$$

Since \mathbf{p} belongs to the column space of \mathbf{Z} , we must have the matrix equation $\mathbf{Z}\mathbf{x} = \mathbf{p}$ to be consistent. Furthermore, we have $d(\mathbf{p}, \mathbf{v}_1) \leq d(\mathbf{Z}\mathbf{x}, \mathbf{v}_1)$ for all $\mathbf{x} \in \mathbb{R}^4$. So a vector \mathbf{w} such that $\mathbf{Z}\mathbf{x} = \mathbf{w}$ is consistent and $d(\mathbf{w}, \mathbf{v}_1)$ is as small as possible is $\mathbf{p} = \left(\frac{5}{4}, -\frac{3}{4}, -\frac{5}{4}, \frac{7}{4} \right)^T$.

- (iv) Arguing by contradiction, suppose there exist elementary matrices $\mathbf{E}_1, \mathbf{E}_2, \dots, \mathbf{E}_k$, such that $\mathbf{E}_k \cdots \mathbf{E}_1 \mathbf{Y} = \mathbf{Z}$. Then this would imply that \mathbf{Y} is row-equivalent to \mathbf{Z} . In particular, \mathbf{Y} and \mathbf{Z} must have the same row space, and hence the same rank. But from part (iii), we see that $\text{rank}(\mathbf{Z}) = 3 \neq 2 = \text{rank}(\mathbf{Y})$, which is a contradiction. So there do not exist elementary matrices $\mathbf{E}_1, \mathbf{E}_2, \dots, \mathbf{E}_k$, such that $\mathbf{E}_k \cdots \mathbf{E}_1 \mathbf{Y} = \mathbf{Z}$.
- (v) Since $\mathbf{Y}\mathbf{u}_1, \mathbf{Y}\mathbf{u}_2, \mathbf{Y}\mathbf{u}_3, \mathbf{Y}\mathbf{u}_4$ all belong to the column space of \mathbf{Y} , it follows that U is a vector subspace of the column space of \mathbf{Y} , and hence $\dim(\text{span}(U)) \leq \text{rank}(\mathbf{Y}) = 2$. As $\{(2, 2, 4, 0)^T, (-4, -1, -5, -1)^T\}$ is clearly linearly independent, and $\mathbf{Y}\mathbf{e}_1 = (2, 2, 4, 0)^T$, $\mathbf{Y}\mathbf{e}_2 = (-4, -1, -5, -1)^T$, we see that by setting $\mathbf{u}_1 = \mathbf{e}_1$, $\mathbf{u}_2 = \mathbf{e}_2$, and $\mathbf{u}_3 = \mathbf{u}_4 = \mathbf{0}$, we have $\{\mathbf{Y}\mathbf{u}_1, \mathbf{Y}\mathbf{u}_2\}$ to form a basis for $\text{span}(U)$. So the largest possible value of $\dim(\text{span}(U))$ is 2.