

MA 106 : Linear Algebra

Tutorial 5

Soumya Chatterjee

IIT Bombay

7th April 2021

Question 1(i)

Find all eigenvalues, and their geometric as well as algebraic multiplicities of the following matrices. Are they diagonalizable? If so, find invertible \mathbf{P} such that $\mathbf{P}^{-1}\mathbf{A}\mathbf{P}$ is a diagonal matrix.

$$\mathbf{A} := \begin{bmatrix} 5 & -1 \\ 1 & 3 \end{bmatrix}$$

- $p_{\mathbf{A}}(t) = \det \begin{bmatrix} 5-t & -1 \\ 1 & 3-t \end{bmatrix}$
- $= (5-t)(3-t) + 1 = t^2 - 8t + 16$
- $= (t-4)^2$
- $\mathbf{A} - 4\mathbf{I} = \det \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix} \rightarrow \text{rank}(\mathbf{A} - 4\mathbf{I}) = 1$
- Eigenvalue: 4 with algebraic multiplicity 2 and geometric multiplicity = nullity($\mathbf{A} - 4\mathbf{I}$) = $2 - 1 = 1$
- Not diagonalizable

Question 1(ii)

$$\mathbf{A} := \begin{bmatrix} 3 & 2 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 2 & -1 & 0 \\ 0 & 0 & 0 & 1/2 \end{bmatrix}$$

- $p_{\mathbf{A}}(t) = \det \begin{bmatrix} 3-t & 2 & 1 & 0 \\ 0 & 1-t & 0 & 1 \\ 0 & 2 & -1-t & 0 \\ 0 & 0 & 0 & 1/2-t \end{bmatrix}$
- $= (3-t) \begin{bmatrix} 1-t & 0 & 1 \\ 2 & -1-t & 0 \\ 0 & 0 & 1/2-t \end{bmatrix}$
- $= (3-t)(1/2-t) \begin{bmatrix} 1-t & 0 \\ 2 & -1-t \end{bmatrix} = (3-t)(1/2-t)(1-t)(-1-t)$
- Eigenvalues: $\{3, 1/2, 1, -1\}$ all with algebraic and geometric multiplicity 1
- Diagonalizable

- $\mathbf{A} = \begin{bmatrix} 3 & 2 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 2 & -1 & 0 \\ 0 & 0 & 0 & 1/2 \end{bmatrix}$ has eigenvalues: $\{3, 1/2, 1, -1\}$

- For $\lambda = 3$,

▶ $\mathbf{A} - 3\mathbf{I} = \begin{bmatrix} 0 & 2 & 1 & 0 \\ 0 & -2 & 0 & 1 \\ 0 & 2 & -4 & 0 \\ 0 & 0 & 0 & -5/2 \end{bmatrix} \rightarrow \begin{bmatrix} 0 & 2 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & -5 & 0 \\ 0 & 0 & 0 & -5/2 \end{bmatrix} \rightarrow$

$\begin{bmatrix} 0 & 2 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & -5/2 \\ 0 & 0 & 0 & 0 \end{bmatrix}$. Basic solution = $[1 \ 0 \ 0 \ 0]^T$

- Similarly, check that for $\lambda = 1/2$, $[8/3 \ -2 \ -8/3 \ 1]^T$;
 $\lambda = 1$, $[-3/2 \ 1 \ 1 \ 0]^T$; $\lambda = -1$, $[-1/4 \ 0 \ 1 \ 0]^T$

- $\mathbf{P} = \begin{bmatrix} 1 & 8/3 & -3/2 & -1/4 \\ 0 & -2 & 1 & 0 \\ 0 & -8/3 & 1 & 1 \\ 0 & 1 & 0 & 0 \end{bmatrix}$

Question 1(iii)

$$\mathbf{A} := \begin{bmatrix} 2 & 1 & 0 \\ 0 & 2 & 1 \\ 0 & 0 & 2 \end{bmatrix}$$

- $p_{\mathbf{A}}(t) = \det \begin{bmatrix} 2-t & 1 & 0 \\ 0 & 2-t & 1 \\ 0 & 0 & 2-t \end{bmatrix} = (2-t)^3$
- $\mathbf{A} - 2\mathbf{I} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \Rightarrow \text{rank}(\mathbf{A} - 2\mathbf{I}) = 2$
- Eigenvalue: 4 with algebraic multiplicity 2 and geometric multiplicity = nullity($\mathbf{A} - 2\mathbf{I}$) = $3 - 2 = 1$
- Not diagonalizable

Question 2

Let $\mathbf{A} := \begin{bmatrix} 2 & a & b \\ 0 & 1 & c \\ 0 & 0 & 2 \end{bmatrix}$. Find a necessary and sufficient condition on a, b, c for \mathbf{A} to be diagonalizable.

- $p_{\mathbf{A}}(t) = \det \begin{bmatrix} 2-t & a & b \\ 0 & 1-t & c \\ 0 & 0 & 2-t \end{bmatrix} = (2-t)^2(1-t)$
- For \mathbf{A} to be diagonalizable, we need the algebraic and geometric multiplicities of the eigenvalues to match
- For $\lambda = 1$, this is always true (why?).
- For $\lambda = 2$, we need its geometric multiplicity to be 2, or $\text{nullity}(\mathbf{A} - 2\mathbf{I}) = 2$, or $\text{rank}(\mathbf{A} - 2\mathbf{I}) = 1$
- $\mathbf{A} - 2\mathbf{I} = \begin{bmatrix} 0 & a & b \\ 0 & -1 & c \\ 0 & 0 & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 0 & 1 & -c \\ 0 & a & b \\ 0 & 0 & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 0 & 1 & -c \\ 0 & 0 & b+ac \\ 0 & 0 & 0 \end{bmatrix}$
- So, we need $b + ac = 0$

- $\mathbf{A} - \mathbf{I} = \begin{bmatrix} 1 & a & b \\ 0 & 0 & c \\ 0 & 0 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & a & b \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & a & b \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$

- $\text{rank}(\mathbf{A} - \mathbf{I}) = 2 \Rightarrow \text{nullity}(\mathbf{A} - \mathbf{I}) = 1 \Rightarrow \lambda = 1$ has geometric multiplicity 1

Question 4

Let $\lambda \in \mathbb{K}$. Show that λ is an eigenvalue of \mathbf{A} if and only if $\bar{\lambda}$ is an eigenvalue of \mathbf{A}^* , but their eigenvectors can be very different

- $\det \mathbf{A}^* = \det \bar{\mathbf{A}}^\top = \det \bar{\mathbf{A}} = \overline{\det \mathbf{A}}$ (Prove by induction)
- Outline of proof:
 - ▶ Base case: $\det [x]^* = \det [\bar{x}] = \bar{x} = \overline{\det [x]}$
 - ▶ Let $\mathbf{B} = \bar{\mathbf{A}}$ and let N_{ij} be the minors of \mathbf{A}
 - ▶ $\det \mathbf{B} = \sum_n (-1)^{1+i} b_{1i} N_{1i}$
 - ▶ $= \sum_n (-1)^{1+i} \overline{a_{1i}} \overline{M_{1i}}$ (minors are determinants of a smaller size)
 - ▶ $= \sum_n (-1)^{1+i} \overline{a_{1i} M_{1i}} = \overline{\sum_n (-1)^{1+i} a_{1i} M_{1i}} = \overline{\det \mathbf{A}}$
- Thus we have $\det \mathbf{A}^* = \overline{\det \mathbf{A}}$
- We have $\det(\mathbf{A} - \lambda \mathbf{I}) = \overline{\det(\mathbf{A} - \lambda \mathbf{I})} = \overline{\det((\mathbf{A} - \lambda \mathbf{I})^*)}$
 $= \det(\mathbf{A}^* - \bar{\lambda} \mathbf{I})$
- Thus $\det(\mathbf{A} - \lambda \mathbf{I}) = 0 \iff \det(\mathbf{A}^* - \bar{\lambda} \mathbf{I}) = 0$

Question 5

Let $\mathbf{A} \in \mathbb{K}^{n \times n}$. Show that 0 is an eigenvalue of \mathbf{A} if and only if 0 is an eigenvalue of $\mathbf{A}^* \mathbf{A}$, and its geometric multiplicity is the same.

- (\implies)
 - ▶ 0 is an eigenvalue of \mathbf{A}
 - ▶ $\mathbf{Ax} = 0 \implies \mathbf{A}^* \mathbf{Ax} = 0$
 - ▶ 0 is an eigenvalue of $\mathbf{A}^* \mathbf{A}$
- (\impliedby)
 - ▶ 0 is an eigenvalue of $\mathbf{A}^* \mathbf{A}$
 - ▶ $\mathbf{A}^* \mathbf{Ax} = 0 \implies \mathbf{x}^* \mathbf{A}^* \mathbf{Ax} = 0$
 - ▶ $(\mathbf{Ax})^* \mathbf{Ax} = 0 \implies \mathbf{Ax} = 0$ Why?
 - ▶ 0 is an eigenvalue of \mathbf{A}
- Thus we have shown that 0 is an eigenvalue of \mathbf{A} if and only if 0 is an eigenvalue of $\mathbf{A}^* \mathbf{A}$

Let $\mathbf{A} \in \mathbb{K}^{n \times n}$. Show that 0 is an eigenvalue of \mathbf{A} if and only if 0 is an eigenvalue of $\mathbf{A}^* \mathbf{A}$, and its geometric multiplicity is the same.

- Geometric multiplicity = $\dim \mathcal{N}(\mathbf{A} - \lambda \mathbf{I}) = \dim \mathcal{N}(\mathbf{A})$ (for $\lambda = 0$)
- Thus to prove equal geometric multiplicity of 0, we need to show that $\dim \mathcal{N}(\mathbf{A}) = \dim \mathcal{N}(\mathbf{A}^* \mathbf{A})$
- We instead prove a stronger statement: $\mathcal{N}(\mathbf{A}) = \mathcal{N}(\mathbf{A}^* \mathbf{A})$
 - ▶ $\mathbf{x} \in \mathcal{N}(\mathbf{A}) \Leftrightarrow \mathbf{A}\mathbf{x} = 0 \iff \mathbf{A}^* \mathbf{A}\mathbf{x} = 0 \Leftrightarrow \mathbf{x} \in \mathcal{N}(\mathbf{A}^* \mathbf{A})$
 - ▶ Above statement follows from the previous part
 - ▶ $\mathbf{x} \in \mathcal{N}(\mathbf{A}) \Leftrightarrow \mathbf{x} \in \mathcal{N}(\mathbf{A}^* \mathbf{A}) \implies \mathcal{N}(\mathbf{A}) = \mathcal{N}(\mathbf{A}^* \mathbf{A})$

Deduce $\text{rank } \mathbf{A}^* \mathbf{A} = \text{rank } \mathbf{A}$

- ▶ We have shown that $\mathcal{N}(\mathbf{A}) = \mathcal{N}(\mathbf{A}^* \mathbf{A})$
- ▶ $\dim \mathcal{N}(\mathbf{A}) = \dim \mathcal{N}(\mathbf{A}^* \mathbf{A}) \Rightarrow \text{nullity}(\mathbf{A}) = \text{nullity}(\mathbf{A}^* \mathbf{A})$
- ▶ By rank-nullity theorem, $\text{rank } \mathbf{A}^* \mathbf{A} = \text{rank } \mathbf{A}$ follows (both matrices have the same number of columns)

Question 7

A square matrix $\mathbf{A} := [a_{jk}]$ is called **strictly diagonally dominant** if $|a_{jj}| > \sum_{k \neq j} |a_{jk}|$ for each $j = 1, \dots, n$. If \mathbf{A} strictly diagonally dominant, show that \mathbf{A} is invertible.

- Showing \mathbf{A} is invertible is equivalent to showing nullity $\mathbf{A} = 0$
- Or $\mathbf{A}\mathbf{x} = 0$ does not have any non-zero solution or 0 is not an eigenvalue of \mathbf{A}

Recall: Gerschgorin Theorem

Every eigenvalue of $\mathbf{A} \in \mathbb{K}^{n \times n}$ belongs in one of its Gerschgorin disks.

If $r_j := \sum_{k \neq j} |a_{jk}|$, the Gerschgorin disks are given by

$$D(a_{jj}, r_j) := \{a \in \mathbb{K} : |a - a_{jj}| \leq r_j\}$$

- Since for strictly diagonally dominant matrix \mathbf{A} , $|a_{jj}| > r_j$, by Gerschgorin Theorem, none of the disks can contain the origin (distance of center from origin $>$ radius)
- Thus 0 cannot be a eigenvalue of \mathbf{A}

Question 8

Let $\mathbf{A} \in \mathbb{K}^{n \times n}$. Define $\alpha_2 := \max\{\|\mathbf{Ax}\| : \|\mathbf{x}\| = 1\}$,
 $\alpha_\infty := \max\{\sum_{k=1}^n |a_{jk}| : j = 1, \dots, n\}$ and
 $\alpha_1 := \max\{\sum_{j=1}^n |a_{jk}| : k = 1, \dots, n\}$, where $\mathbf{A} := [a_{jk}]$.
Show that $|\lambda| \leq \min\{\alpha_2, \alpha_\infty, \alpha_1\}$ for every eigenvalue λ .

- $|\lambda| \leq \min\{\alpha_2, \alpha_\infty, \alpha_1\} \iff |\lambda| \leq \alpha_2$ and $|\lambda| \leq \alpha_\infty$ and $|\lambda| \leq \alpha_1$
- First we show that $|\lambda| \leq \alpha_2$
 - ▶ Let λ_{\max} be the eigenvalue with maximum absolute value *i.e.*
 $|\lambda_{\max}| \geq |\lambda_i| \forall i \in \{1, 2, \dots, n\}$
 - ▶ Thus there exists a vector \mathbf{x} with $\|\mathbf{x}\| = 1$ such that $\mathbf{Ax} = \lambda_{\max}\mathbf{x}$
 - ▶ $\|\mathbf{Ax}\| = \|\lambda_{\max}\mathbf{x}\| = |\lambda_{\max}|\|\mathbf{x}\| = |\lambda_{\max}|$
 - ▶ Which gives us $\alpha_2 = \max\{\|\mathbf{Ax}\| : \|\mathbf{x}\| = 1\} \geq |\lambda_{\max}|$ (Why?)
 - ▶ $|\lambda_{\max}| \leq \alpha_2 \Rightarrow |\lambda_i| \leq \alpha_2$ for all i (Why?)
- Next we show that $|\lambda| \leq \alpha_\infty$
 - ▶ Let j' be the value of j at which $\sum_k |a_{jk}|$ attains its maximum
 - ▶ By Gershgorin Theorem, $|\lambda - a_{j'j'}| \leq \sum_{k, k \neq j'} |a_{j'k}|$
 - ▶ $|\lambda| - |a_{j'j'}| \leq |\lambda - a_{j'j'}| \leq \sum_{k, k \neq j'} |a_{j'k}|$ (Triangle Inequality)
 - ▶ $|\lambda| \leq \sum_{k, k \neq j'} |a_{j'k}| + |a_{j'j'}| = \alpha_\infty$

Let $\mathbf{A} \in \mathbb{K}^{n \times n}$. Define $\alpha_2 := \max\{\|\mathbf{A}\mathbf{x}\| : \|\mathbf{x}\| = 1\}$,
 $\alpha_\infty := \max\{\sum_{k=1}^n |a_{jk}| : j = 1, \dots, n\}$ and
 $\alpha_1 := \max\{\sum_{j=1}^n |a_{jk}| : k = 1, \dots, n\}$, where $\mathbf{A} := [a_{jk}]$.
Show that $|\lambda| \leq \min\{\alpha_2, \alpha_\infty, \alpha_1\}$ for every eigenvalue λ .

- Finally, we show that $|\lambda| \leq \alpha_1$
 - ▶ Observe that $|\lambda| \leq \alpha_\infty$ is the maximum row sum and $|\lambda| \leq \alpha_1$ is the maximum column sum
 - ▶ If we show that \mathbf{A} and \mathbf{A}^\top have the same eigenvalues, we can apply Gershgorin Theorem as in the previous part and be done
 - ▶ $\det(\mathbf{A} - \lambda\mathbf{I}) = \det((\mathbf{A} - \lambda\mathbf{I})^\top) = \det(\mathbf{A}^\top - \lambda\mathbf{I})$
 - ▶ $\det(\mathbf{A} - \lambda\mathbf{I}) = 0 \Leftrightarrow \det(\mathbf{A}^\top - \lambda\mathbf{I}) = 0$ or both \mathbf{A} and \mathbf{A}^\top have the same eigenvalues
 - ▶ Can you complete the proof?

Question 9

Let $\mathbf{x}, \mathbf{y} \in \mathbb{K}^{n \times 1}$. Prove $\|\mathbf{x} + \mathbf{y}\|^2 + \|\mathbf{x} - \mathbf{y}\|^2 = 2\|\mathbf{x}\|^2 + 2\|\mathbf{y}\|^2$.

- $\|\mathbf{x} + \mathbf{y}\|^2 + \|\mathbf{x} - \mathbf{y}\|^2 = (\mathbf{x} + \mathbf{y})^*(\mathbf{x} + \mathbf{y}) + (\mathbf{x} - \mathbf{y})^*(\mathbf{x} - \mathbf{y})$
- $= (\mathbf{x}^* + \mathbf{y}^*)(\mathbf{x} + \mathbf{y}) + (\mathbf{x}^* - \mathbf{y}^*)(\mathbf{x} - \mathbf{y})$
- $= \mathbf{x}^*\mathbf{x} + \mathbf{y}^*\mathbf{y} + \mathbf{x}^*\mathbf{y} + \mathbf{y}^*\mathbf{x} + \mathbf{x}^*\mathbf{x} + \mathbf{y}^*\mathbf{y} - \mathbf{x}^*\mathbf{y} - \mathbf{y}^*\mathbf{x}$
- $= 2\mathbf{x}^*\mathbf{x} + 2\mathbf{y}^*\mathbf{y} = 2\|\mathbf{x}\|^2 + 2\|\mathbf{y}\|^2$

If \mathbf{x} and \mathbf{y} are nonzero, prove $\|\mathbf{x} - \mathbf{y}\|^2 = \|\mathbf{x}\|^2 + \|\mathbf{y}\|^2 - 2\|\mathbf{x}\|\|\mathbf{y}\|\cos\theta$, where $\theta \in [0, \pi]$ is defined to be $\cos^{-1}(\Re\langle\mathbf{x}, \mathbf{y}\rangle/\|\mathbf{x}\|\|\mathbf{y}\|)$.

- $\|\mathbf{x} - \mathbf{y}\|^2 = (\mathbf{x} - \mathbf{y})^*(\mathbf{x} - \mathbf{y}) = \mathbf{x}^*\mathbf{x} + \mathbf{y}^*\mathbf{y} - \mathbf{x}^*\mathbf{y} - \mathbf{y}^*\mathbf{x}$
- $= \mathbf{x}^*\mathbf{x} + \mathbf{y}^*\mathbf{y} - \langle\mathbf{x}, \mathbf{y}\rangle - \langle\mathbf{y}, \mathbf{x}\rangle = \mathbf{x}^*\mathbf{x} + \mathbf{y}^*\mathbf{y} - \langle\mathbf{x}, \mathbf{y}\rangle - \overline{\langle\mathbf{x}, \mathbf{y}\rangle}$
- $= \mathbf{x}^*\mathbf{x} + \mathbf{y}^*\mathbf{y} - 2\Re(\mathbf{x}^*\mathbf{y}) = \mathbf{x}^*\mathbf{x} + \mathbf{y}^*\mathbf{y} - 2\Re\langle\mathbf{x}, \mathbf{y}\rangle$
- $= \mathbf{x}^*\mathbf{x} + \mathbf{y}^*\mathbf{y} - 2\|\mathbf{x}\|\|\mathbf{y}\|\Re\langle\mathbf{x}, \mathbf{y}\rangle/\|\mathbf{x}\|\|\mathbf{y}\|$
- Now, $\frac{|\Re\langle\mathbf{x}, \mathbf{y}\rangle|}{\|\mathbf{x}\|\|\mathbf{y}\|} \leq \frac{|\langle\mathbf{x}, \mathbf{y}\rangle|}{\|\mathbf{x}\|\|\mathbf{y}\|} \leq 1$ (Schwarz inequality)
- $\Rightarrow -1 \leq \frac{\Re\langle\mathbf{x}, \mathbf{y}\rangle}{\|\mathbf{x}\|\|\mathbf{y}\|} \leq 1$. So, we can conclude that $\theta \in [0, \pi]$ exist

$$\overline{\langle\mathbf{x}, \mathbf{y}\rangle} = \overline{\mathbf{x}^*\mathbf{y}} = \overline{\mathbf{x}^*}\overline{\mathbf{y}} = \overline{\mathbf{x}}^T\overline{\mathbf{y}} = \overline{\mathbf{x}}^T\overline{\mathbf{y}} = \overline{\mathbf{y}}^T\mathbf{x} = \mathbf{y}^*\mathbf{x} = \langle\mathbf{y}, \mathbf{x}\rangle$$

Question 3 and 6

- See the OneNote notebook in MS Teams

Tutorial 6 Question 1

Orthonormalize the following ordered subsets of $\mathbb{K}^{4 \times 1}$

(i) $(\mathbf{e}_1, \mathbf{e}_1 + \mathbf{e}_2, \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3, \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 + \mathbf{e}_4)$

▶ $\mathbf{y}_1 = \mathbf{e}_1$

▶ $\mathbf{y}_2 = \mathbf{e}_1 + \mathbf{e}_2 - \frac{\langle \mathbf{e}_1, \mathbf{e}_1 + \mathbf{e}_2 \rangle}{\langle \mathbf{e}_1, \mathbf{e}_1 \rangle} \mathbf{e}_1 = \mathbf{e}_2$

▶ $\mathbf{y}_3 = \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 - \frac{\langle \mathbf{e}_1, \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 \rangle}{\langle \mathbf{e}_1, \mathbf{e}_1 \rangle} \mathbf{e}_1 - \frac{\langle \mathbf{e}_2, \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 \rangle}{\langle \mathbf{e}_2, \mathbf{e}_2 \rangle} \mathbf{e}_2 = \mathbf{e}_3$

▶ $\mathbf{y}_4 = \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 + \mathbf{e}_4 - \frac{\langle \mathbf{e}_1, \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 + \mathbf{e}_4 \rangle}{\langle \mathbf{e}_1, \mathbf{e}_1 \rangle} \mathbf{e}_1 - \frac{\langle \mathbf{e}_2, \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 + \mathbf{e}_4 \rangle}{\langle \mathbf{e}_2, \mathbf{e}_2 \rangle} \mathbf{e}_2 - \frac{\langle \mathbf{e}_3, \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 + \mathbf{e}_4 \rangle}{\langle \mathbf{e}_3, \mathbf{e}_3 \rangle} \mathbf{e}_3 = \mathbf{e}_4$

(ii) $(\mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 + \mathbf{e}_4, -\mathbf{e}_1 + \mathbf{e}_2, -\mathbf{e}_1 + \mathbf{e}_3, -\mathbf{e}_1 + \mathbf{e}_4)$

▶ $\mathbf{y}_1 = \mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 + \mathbf{e}_4$

▶ $\mathbf{y}_2 = -\mathbf{e}_1 + \mathbf{e}_2 - \frac{\langle \mathbf{y}_1, -\mathbf{e}_1 + \mathbf{e}_2 \rangle}{\langle \mathbf{y}_1, \mathbf{y}_1 \rangle} \mathbf{y}_1 = -\mathbf{e}_1 + \mathbf{e}_2$

▶ $\mathbf{y}_3 = -\mathbf{e}_1 + \mathbf{e}_3 - 0 \cdot \mathbf{y}_1 - \frac{\langle -\mathbf{e}_1 + \mathbf{e}_3, -\mathbf{e}_1 + \mathbf{e}_2 \rangle}{\langle -\mathbf{e}_1 + \mathbf{e}_2, -\mathbf{e}_1 + \mathbf{e}_2 \rangle} (-\mathbf{e}_1 + \mathbf{e}_2) = -\mathbf{e}_1 + \mathbf{e}_3 - \frac{-\mathbf{e}_1 + \mathbf{e}_2}{2} = -\frac{1}{2} \mathbf{e}_1 - \frac{1}{2} \mathbf{e}_2 + \mathbf{e}_3$

▶ $\mathbf{y}_4 = -\mathbf{e}_1 + \mathbf{e}_4 - 0 \cdot \mathbf{y}_1 - \frac{\langle -\mathbf{e}_1 + \mathbf{e}_4, \mathbf{y}_2 \rangle}{\langle \mathbf{y}_2, \mathbf{y}_2 \rangle} \mathbf{y}_2 - \frac{\langle -\mathbf{e}_1 + \mathbf{e}_4, \mathbf{y}_3 \rangle}{\langle \mathbf{y}_3, \mathbf{y}_3 \rangle} \mathbf{y}_3 = -\mathbf{e}_1 + \mathbf{e}_4 - \frac{1}{2} \mathbf{y}_2 - \frac{1/2}{3/2} \mathbf{y}_3 = -\frac{1}{3} \mathbf{e}_1 - \frac{1}{3} \mathbf{e}_2 - \frac{1}{3} \mathbf{e}_3 + \mathbf{e}_4$

▶ $(\frac{\mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3 + \mathbf{e}_4}{2}, \frac{-\mathbf{e}_1 + \mathbf{e}_2}{\sqrt{2}}, -\frac{\sqrt{3}}{2\sqrt{2}} \mathbf{e}_1 - \frac{\sqrt{3}}{2\sqrt{2}} \mathbf{e}_2 + \frac{\sqrt{3}}{\sqrt{2}} \mathbf{e}_3, -\frac{\mathbf{e}_1 + \mathbf{e}_2 + \mathbf{e}_3}{2\sqrt{3}} + \frac{\sqrt{3}}{2} \mathbf{e}_4)$

Tutorial 6 Question 2

Use the Gram-Schmidt Orthogonalization Process to orthonormalize the ordered subset

$$\left(\begin{bmatrix} 1 & -1 & 2 & 0 \end{bmatrix}^T, \begin{bmatrix} 1 & 1 & 2 & 0 \end{bmatrix}^T, \begin{bmatrix} 3 & 0 & 0 & 1 \end{bmatrix}^T \right)$$

and obtain an ordered orthonormal set $(\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3)$.

- $\mathbf{y}_1 = \begin{bmatrix} 1 & -1 & 2 & 0 \end{bmatrix}^T$
- $\mathbf{y}_2 = \begin{bmatrix} 1 & 1 & 2 & 0 \end{bmatrix}^T - \frac{4}{6} \begin{bmatrix} 1 & -1 & 2 & 0 \end{bmatrix}^T = \begin{bmatrix} \frac{1}{3} & \frac{5}{3} & \frac{2}{3} & 0 \end{bmatrix}^T$
- $\mathbf{y}_3 = \begin{bmatrix} 3 & 0 & 0 & 1 \end{bmatrix}^T - \frac{3}{6} \begin{bmatrix} 1 & -1 & 2 & 0 \end{bmatrix}^T - \frac{1}{10/3} \begin{bmatrix} \frac{1}{3} & \frac{5}{3} & \frac{2}{3} & 0 \end{bmatrix}^T = \begin{bmatrix} \frac{12}{5} & 0 & -\frac{6}{5} & 1 \end{bmatrix}^T$

- $\mathbf{u}_1 = \frac{\mathbf{y}_1}{\|\mathbf{y}_1\|} = \left[\frac{\sqrt{6}}{6} \quad -\frac{\sqrt{6}}{6} \quad \frac{\sqrt{6}}{3} \quad 0 \right]^T$
- $\mathbf{u}_2 = \frac{\mathbf{y}_2}{\|\mathbf{y}_2\|} = \left[\frac{\sqrt{30}}{30} \quad \frac{\sqrt{30}}{6} \quad \frac{\sqrt{30}}{15} \quad 0 \right]^T$
- $\mathbf{u}_3 = \frac{\mathbf{y}_3}{\|\mathbf{y}_3\|} = \left[\frac{12\sqrt{205}}{205} \quad 0 \quad -\frac{6\sqrt{205}}{205} \quad \frac{\sqrt{205}}{41} \right]^T$

Find \mathbf{u}_4 such that $(\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3, \mathbf{u}_4)$ is an ordered orthonormal basis for $\mathbb{K}^{4 \times 1}$

- We want to \mathbf{u}_4 such that $\langle \mathbf{u}_1, \mathbf{u}_4 \rangle = \langle \mathbf{u}_2, \mathbf{u}_4 \rangle = \langle \mathbf{u}_3, \mathbf{u}_4 \rangle = 0$
- Find the basic solution of (or directly use \mathbf{y}_i instead of \mathbf{u}_i),

$$\begin{bmatrix} \frac{\sqrt{6}}{6} & -\frac{\sqrt{6}}{6} & \frac{\sqrt{6}}{3} & 0 \\ \frac{\sqrt{30}}{30} & \frac{\sqrt{30}}{6} & \frac{\sqrt{30}}{15} & 0 \\ \frac{12\sqrt{205}}{205} & 0 & -\frac{6\sqrt{205}}{205} & \frac{\sqrt{205}}{41} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

- It turns out to be $\mathbf{y}_4 = \left[-\frac{1}{3} \quad 0 \quad \frac{1}{6} \quad 1 \right]^T$
- $\mathbf{u}_4 = \left[-\frac{2}{\sqrt{41}} \quad 0 \quad \frac{1}{\sqrt{41}} \quad \frac{6}{\sqrt{41}} \right]^T$

Express the vector $\mathbf{x} = [1 \quad -1 \quad 1 \quad -1]^T$ as a linear combination of these four basis vectors.

$$\bullet \mathbf{u}_1 = \left[\frac{\sqrt{6}}{6} \quad -\frac{\sqrt{6}}{6} \quad \frac{\sqrt{6}}{3} \quad 0 \right]^T$$

$$\bullet \mathbf{u}_2 = \left[\frac{\sqrt{30}}{30} \quad \frac{\sqrt{30}}{6} \quad \frac{\sqrt{30}}{15} \quad 0 \right]^T$$

$$\bullet \mathbf{u}_3 = \left[\frac{12\sqrt{205}}{205} \quad 0 \quad -\frac{6\sqrt{205}}{205} \quad \frac{\sqrt{205}}{41} \right]^T$$

$$\bullet \mathbf{u}_4 = \left[-\frac{2\sqrt{41}}{41} \quad 0 \quad \frac{\sqrt{41}}{41} \quad \frac{6\sqrt{41}}{41} \right]^T$$

$$\bullet \mathbf{x} = P_{\mathbf{u}_1}(\mathbf{x}) + P_{\mathbf{u}_2}(\mathbf{x}) + P_{\mathbf{u}_3}(\mathbf{x}) + P_{\mathbf{u}_4}(\mathbf{x})$$

$$\bullet \mathbf{x} = \langle \mathbf{u}_1, \mathbf{x} \rangle \mathbf{u}_1 + \langle \mathbf{u}_2, \mathbf{x} \rangle \mathbf{u}_2 + \langle \mathbf{u}_3, \mathbf{x} \rangle \mathbf{u}_3 + \langle \mathbf{u}_4, \mathbf{x} \rangle \mathbf{u}_4 \quad (\langle \mathbf{u}, \mathbf{u} \rangle = 1)$$

$$\bullet \mathbf{x} = \frac{2\sqrt{6}}{3} \mathbf{u}_1 - \frac{1\sqrt{30}}{15} \mathbf{u}_2 + \frac{\sqrt{205}}{205} \mathbf{u}_3 - \frac{7\sqrt{41}}{41} \mathbf{u}_4$$