

Question 1.

1. According to the properties of determinants, the determinant of any square matrix A can be represented as $\det(A) = \frac{\prod \text{diag}(B)}{d}$, where B is the REF of A after Gaussian elimination, and d is the product of the scalars by applying Gaussian elimination:

① Swapping: multiplies $\det(A)$ by -1

② Multiplying a row by a scalar: multiplies $\det(A)$ by the same scalar

③ Adding to one row a scalar multiple of another: doesn't change $\det(A)$

Then, $\det(A) = \frac{1}{d} \prod \text{diag}(B)$, where $d \neq 0$.

After applying Gaussian elimination on a $n \times n$ permutation matrix P, we have the REF of P and denote it as PREF. From the structure of P, we induce that $\text{PREF} = I_n$, then $\text{diag}(\text{PREF}) = 1 \times 1 \times \dots \times 1 = 1$. Then, $\det(P) = \frac{1}{d} \neq 0$, where d is a non-zero scalar produced by Gaussian elimination of P.

Thus, $\det(P) \neq 0 \Rightarrow P$ is always invertible.

2. Suppose p_{ij} is the entry of P, according to the definition of permutation matrices, we have $P_{1,j_1} = P_{2,j_2} = \dots = P_{n,j_n} = 1$, where $j_1 \neq j_2 \neq \dots \neq j_n$, and other entries are all 0s.

After transpose, P^T has entries that $P_{j_1,1}^T = P_{j_2,2}^T = \dots = P_{j_n,n}^T = 1$, where $j_1 \neq j_2 \neq \dots \neq j_n$, and other entries are all 0s.

Thus, based on the definition of permutation matrices, we have P^T is a permutation matrix.

Question 2.

1. Suppose x_i can be written as $x_i = c_1 x_1 + \dots + c_{i-1} x_{i-1} + c_{i+1} x_{i+1} + \dots + c_n x_n$, ..

where $c_1, \dots, c_{i-1}, c_{i+1}, \dots, c_n$ can not all be 0s.

Then $Ax_i = \lambda_1 x_1$

$$\begin{aligned} &= \lambda_1 (c_1 x_1 + \dots + c_{i-1} x_{i-1} + c_{i+1} x_{i+1} + \dots + c_n x_n) \\ &= \lambda_1 c_1 x_1 + \dots + \lambda_1 c_{i-1} x_{i-1} + \lambda_1 c_{i+1} x_{i+1} + \dots + \lambda_1 c_n x_n \quad \textcircled{1} \end{aligned}$$

Also, $Ax_i = A(c_1 x_1 + \dots + c_{i-1} x_{i-1} + c_{i+1} x_{i+1} + \dots + c_n x_n)$

$$\begin{aligned} &= c_1 Ax_1 + \dots + c_{i-1} Ax_{i-1} + c_{i+1} Ax_{i+1} + \dots + c_n Ax_n \\ &= c_1 \lambda_1 x_1 + \dots + c_{i-1} \lambda_{i-1} x_{i-1} + c_{i+1} \lambda_{i+1} x_{i+1} + \dots + c_n \lambda_n x_n \quad \textcircled{2} \end{aligned}$$

$$\textcircled{1} - \textcircled{2} = 0 \Rightarrow c_1 (\lambda_1 - \lambda_1) x_1 + \dots + c_{i-1} (\lambda_1 - \lambda_{i-1}) x_{i-1} + c_{i+1} (\lambda_{i+1} - \lambda_{i+1}) x_{i+1} + \dots + c_n (\lambda_n - \lambda_n) x_n = 0$$

From assumptions, we know that $\lambda_1 \neq \lambda_2 \neq \dots \neq \lambda_n$, which implies $\lambda_i - \lambda_k \neq 0$, $k = 1, \dots, i-1, i+1, \dots, n$. We also know that x_k , $k = 1, \dots, i-1, i+1, \dots, n$, is not a zero vector.

It implies that $c_1 = c_2 = \dots = c_{i-1} = c_{i+1} = \dots = c_n = 0$, which conflicts with assumptions.

Thus, $\{x_1, \dots, x_n\}$ are linear independent vectors. Q.E.D.

2. We know that $X \in \mathbb{R}^{n \times n}$, and applying Gaussian elimination only changes the scalar product of $\det(X)$. Following the steps, let $A = B - \lambda I \in \mathbb{R}^{n \times n}$. After Gaussian elimination, the REF of A , denoted as A_{REF} , has at most n non-zero entries on its diagonal i.e. full rank, then $\det(A_{\text{REF}}) = 0$ if there is at least one zero entry on the diagonal, otherwise, $\det(A_{\text{REF}}) = C \times \prod_{i=1}^n A_{\text{REF}ii}$, where $C \in \mathbb{R}$, $A_{\text{REF}ii} \in \mathbb{R}$ is the diagonal entry. We know that $\prod \text{diag}(A_{\text{REF}})$ can be represented as $(a_{11} - \lambda) \cdot (a_{22} - \lambda) \cdots (a_{nn} - \lambda)$, ① When ① = 0, it has at most n roots. (fundamental theorem of algebra) i.e. at most n distinct eigenvalues $\lambda_1 \cdots \lambda_n$. Q.E.D.

Question 3

1. Suppose we have a lower triangular matrix $A \in \mathbb{R}^{n \times n}$, $n \in \mathbb{Z}^+$. From the definition of lower triangular matrices, we know that for $1 < i < j < n$, $a_{ij} = 0$, where a_{ij} is an entry of A , and for $1 < j \leq i < n$, $a_{ij} \in \mathbb{R}$.

Let $B \in \mathbb{R}^{n \times n}$, $b_{ij} \in \mathbb{R}$. Then, we have $C = AB \in \mathbb{R}^{n \times n}$, where $c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}$. For $k < j$, $c_{ij} = \sum_{k=1}^n 0 \cdot b_{kj} = 0$; For $k \geq j$, $c_{ij} = \sum_{k=j}^n a_{ik} b_{kj} \in \mathbb{R}$.

Thus, according to the definition of lower triangular matrices, $C = AB$ is a lower triangular matrix.

If $C = BA \in \mathbb{R}^{n \times n}$, then $c_{ij} = \sum_{k=1}^n b_{ik} a_{kj}$. For $k < j$, $a_{kj} = 0$, then $c_{ij} = 0$. For $k \geq j$, $a_{kj} \in \mathbb{R}$, $c_{ij} \in \mathbb{R}$. Thus, $C = BA$ is also a lower triangular matrix.

In summary, for any lower triangular matrix A , its matrix multiplication is also a lower triangular matrix.

Q.E.D.

2. When $d=1$ $U^1 = [U_{11}]$, then $\det(U^1) = \prod \text{diag}(U^1) = U_{11}$. The base case is true.

Assume for $d=n-1$, $U^{n-1} = \begin{bmatrix} U_{11} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ U_{n-1} & \cdots & U_{n-1}U_{nn} \end{bmatrix} \in \mathbb{R}^{(n-1) \times (n-1)}$, $\det(U^{n-1}) = \prod \text{diag}(U^{n-1}) = U_{11} \times U_{22} \times \cdots \times U_{n-1}$

Then, for $k=n$, $U^n = \begin{bmatrix} U^{n-1} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ U_{n-1} & \cdots & U_{n-1}U_{nn} \end{bmatrix} \in \mathbb{R}^{n \times n}$, $\det(U^n) = U_{nn} \cdot | \cdot | + U_{2n} \cdot | \cdot | + \cdots + U_{nn} \cdot | U^n |$ where $U_{1n} = U_{2n} = \cdots = U_{n-1n} = 0$, $U_{nn} \in \mathbb{R}$.

Then, RHS = $U_{nn} \cdot \det(U^{n-1}) = U_{nn} \cdot \prod \text{diag}(U^{n-1}) = U_{11} \times U_{22} \times \cdots \times U_{nn} = \prod \text{diag}(U^n)$

Thus, the induction holds for $k=1, \dots, n$, i.e. for $U \in \mathbb{R}^{n \times n}$, $\det(U) = \prod \text{diag}(U)$

Q.E.D.

Question 4.

1. Suppose $M = Av_1 = [m_1, \dots, m_n]^T \in \mathbb{R}^{n \times 1}$, where $m_k = \sum_{i=1}^n a_{ki} \cdot v_{1i}$.
 Then $v_2^T M = v_2^T [m_1, \dots, m_n]^T \in \mathbb{R}$, $RHS = \sum_{j=1}^n v_{2j} m_j = \sum_{j=1}^n \sum_{i=1}^n v_{2j} v_{1i} a_{ji}$ ①
 For $N = Av_2$, following the same step, we have $v_1^T N = \sum_{i=1}^n \sum_{j=1}^n v_{1i} v_{2j} a_{ij}$ ②
 From ① and ②, it can be easily showed that $v_2^T M = v_1^T N$
 i.e. $v_2^T A v_1 = v_1^T A v_2$, then $\lambda_1 v_2^T v_1 = \lambda_2 v_1^T v_2$.
 Given the identity $a^T b = b^T a$, we have $\lambda_1 v_1^T v_2 = \lambda_2 v_1^T v_2$.
 Because $\lambda_1 \neq \lambda_2$, then $v_1^T v_2 = 0$, i.e. v_1 and v_2 are orthogonal.
 Q.E.D.

Question 5.

1. Let $Ax = \lambda x$, where x is its eigenvector and λ is the corresponding eigenvalue..

Then, let $Ax = \lambda I_2 x \Rightarrow (A - \lambda I_2)x = 0$, i.e. $\begin{bmatrix} 6-\lambda & 4 \\ 3 & 5-\lambda \end{bmatrix} x = 0$

Suppose $B = A - \lambda I_2$, then we have

$\det(B) = (6-\lambda)(5-\lambda) - 3 \times 4$. When $\det(B) = 0$, we have $\lambda_1 = 2$ and $\lambda_2 = 9$.

2. For $\lambda_1 = 2$. $(A - \lambda_1 I_2)x = 0 \Rightarrow \begin{bmatrix} 4 & 4 \\ 3 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ ①

Solve ①, we have $x_1 = x_2 = 0$

Then $E_{\lambda_1} = \text{span} \left(\begin{pmatrix} 1 \\ 1 \end{pmatrix} \right)$

For $\lambda_2 = 9$. $(A - \lambda_2 I_2)x = 0 \Rightarrow \begin{bmatrix} -3 & 4 \\ 3 & -4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ ②

Solve ②, we have $x_1 = 4n$, $x_2 = 3n$, $n \in \mathbb{R}$, $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 4n \\ 3n \end{bmatrix}$

Then $E_{\lambda_2} = \text{span} \left(\begin{pmatrix} 4 \\ 3 \end{pmatrix} \right)$

3. From the above, we know that $E_{\lambda_1} = \text{span} \left(\begin{pmatrix} 1 \\ 1 \end{pmatrix} \right)$, $E_{\lambda_2} = \text{span} \left(\begin{pmatrix} 4 \\ 3 \end{pmatrix} \right)$, Let $a = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$, $b = \begin{pmatrix} 4 \\ 3 \end{pmatrix}$.

Suppose $A = \begin{bmatrix} a^T \\ b^T \end{bmatrix}$, then $Ax = 0$ has only a trivial solution set, i.e. $x_1 = x_2 = 0$
 which means a and b are independent vectors in \mathbb{R}^2 .

Suppose $ma + nb = \begin{pmatrix} x \\ y \end{pmatrix}$, $x, y \in \mathbb{R}$, then we have $m = \frac{4}{7}x - \frac{3}{7}y$, $n = \frac{x}{7} + \frac{y}{7}$

i.e. a and b span \mathbb{R}^2

Thus, all eigenvectors of A span \mathbb{R}^2 .

4. Following the steps of eigen decomposition. given the eigenvalues and its corresponding eigenvectors we have $\lambda_1 = 2$, $\lambda_2 = 9$, $E_{\lambda_1} = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$, $E_{\lambda_2} = \text{span}(\begin{bmatrix} 4 \\ 1 \end{bmatrix})$
 Then we construct $D = \begin{bmatrix} a & 0 \\ 0 & b \end{bmatrix} \in \mathbb{R}^{2 \times 2}$ as $a = \lambda_1 = 2$, $b = \lambda_2 = 9$, $D = \begin{bmatrix} 2 & 0 \\ 0 & 9 \end{bmatrix}$
 and $P = \begin{bmatrix} 1 & 4 \\ -1 & 1 \end{bmatrix}$, $\det(P) = 1 \times 3 - 4 \times (-1) = 7 \neq 0$, P^{-1} exists

$$5. A^n = (PDP^{-1})^n = PDP^{-1} \times PDP^{-1} \times \cdots \times PDP^{-1} = PD^n P^{-1}$$

$$\text{Thus, } D^n = \begin{bmatrix} 2^n & 0 \\ 0 & 9^n \end{bmatrix} \begin{bmatrix} 2^n & 0 \\ 0 & 9^n \end{bmatrix} \cdots \begin{bmatrix} 2^n & 0 \\ 0 & 9^n \end{bmatrix} = \begin{bmatrix} 2^n & 0 \\ 0 & 9^n \end{bmatrix}$$

$$\text{For } P^{-1}, [P^{-1} | I] = \left[\begin{array}{cc|cc} 1 & 4 & 1 & 0 \\ -1 & 3 & 0 & 1 \end{array} \right]$$

$$\xrightarrow{R_2 \leftarrow R_2 + R_1} \left[\begin{array}{cc|cc} 1 & 4 & 1 & 0 \\ 0 & 7 & 1 & 1 \end{array} \right]$$

$$\xrightarrow{R_2 \leftarrow \frac{1}{7}R_2} \left[\begin{array}{cc|cc} 1 & 4 & 1 & 0 \\ 0 & 1 & \frac{1}{7} & \frac{1}{7} \end{array} \right]$$

$$\xrightarrow{R_1 \leftarrow -4R_2 + R_1} \left[\begin{array}{cc|cc} 1 & 0 & \frac{3}{7} & -\frac{4}{7} \\ 0 & 1 & \frac{1}{7} & \frac{1}{7} \end{array} \right] = [I | P^{-1}] \quad \therefore P^{-1} = \begin{bmatrix} \frac{3}{7} & -\frac{4}{7} \\ \frac{1}{7} & \frac{1}{7} \end{bmatrix}$$

From above, we know that $P = \begin{bmatrix} 1 & 4 \\ -1 & 3 \end{bmatrix}$.

$$\text{Then } A^n = \begin{bmatrix} 1 & 4 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 2^n & 0 \\ 0 & 9^n \end{bmatrix} \begin{bmatrix} \frac{3}{7} & -\frac{4}{7} \\ \frac{1}{7} & \frac{1}{7} \end{bmatrix} = \begin{bmatrix} \frac{3 \times 2^n + 4 \times 9^n}{7} & \frac{-4(2^n - 9^n)}{7} \\ \frac{-3(2^n - 3^{2n})}{7} & \frac{2^{n+2} + 3^{2n+1}}{7} \end{bmatrix}$$

$$= \frac{1}{7} \begin{bmatrix} 3 \times 2^n + 4 \times 9^n & -4(2^n - 9^n) \\ -3(2^n - 3^{2n}) & 2^{n+2} + 3^{2n+1} \end{bmatrix}$$