

بعض الإثباتات لمادة اللينير

① If x_1, x_2 are solutions of $Ax = b$, then $\alpha x_1 + \beta x_2$ is a solution of $Ax = b$ iff $\alpha + \beta = 1$.

Proof \Rightarrow Given $\Rightarrow Ax_1 = b$ and $Ax_2 = b$.

$$\begin{aligned}\text{Then } \Rightarrow A(\alpha x_1 + \beta x_2) &= \alpha(Ax_1) + \beta(Ax_2) \\ &= \alpha b + \beta b = (\alpha + \beta)b = b. \\ \text{iff } \alpha + \beta &= 1\end{aligned}$$

② x_1, x_2 are solutions of a homogeneous linear system $Ax = 0$, then $\alpha x_1 + \beta x_2$ is a solution of $Ax = 0$, $\forall \alpha, \beta \in \mathbb{R}$.

$$\begin{aligned}\text{Proof } \Rightarrow A(\alpha x_1 + \beta x_2) &= \alpha \overset{0}{Ax_1} + \beta \overset{0}{Ax_2} \\ &= 0, \forall \alpha \text{ and } \beta.\end{aligned}$$

⑤ If A is an $m \times n$ matrix. The $A^T A$ and $A A^T$ both symmetric.

$$\begin{aligned}\text{Proof } \Rightarrow B, \bar{A}A &\Rightarrow B^T = (A^T A)^T = A^T (A^T)^T = A^T A \text{ is symmetric.} \\ B &= (A A^T)^T = (A^T)^T A^T = A A^T \text{ is symmetric.}\end{aligned}$$

⑥ If A is symmetric and skew symmetric, then A must be zero matrix.

$$\begin{aligned}\text{Proof } \Rightarrow \alpha A^T &= \alpha A^T = \alpha A \\ \text{since } A &\text{ is symmetric} \\ \therefore \alpha A &\text{ is symmetric.}\end{aligned}$$

[3] Consistency of the linear system:-

A linear system $Ax=b$ is consistent iff b is a linear combination of the columns of A .

Proof \Rightarrow suppose that $Ax=b$ is consistent, so there exist real numbers $\alpha_1, \alpha_2, \dots, \alpha_n$ such that $A \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix} = b$. So,

$$\alpha_1 a_1 + \alpha_2 a_2 + \dots + \alpha_n a_n = b.$$

and so b is a linear combination of the column of A .
conversely,

Suppose that b is a linear combination of the column of A , so there exist real numbers $\alpha_1, \alpha_2, \dots, \alpha_n$ such that $b = \alpha_1 a_1 + \alpha_2 a_2 + \dots + \alpha_n a_n$.

$\Rightarrow b = A \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix}$. So, $\begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix}$ is a solution of $Ax=b$. that is,

$Ax=b$ is consistent.

[4] Let A, B be symmetric matrices. then $H = AB - BA$ is skew symmetric.

Proof $\Rightarrow A^T = A$ and $B^T = B$.

$$\begin{aligned} H^T &= (AB - BA)^T = (AB)^T - (BA)^T \\ &= B^T A^T - A^T B^T \\ &= BA - AB = \boxed{-H} \end{aligned}$$

□ Show that $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ has no inverse (singular matrix).

Proof \Rightarrow if B is any 2×2 matrix, the

$$BA = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} b_{11} & 0 \\ b_{21} & 0 \end{bmatrix} \neq I$$

□ If A and B are nonsingular $n \times n$ matrices, then AB is also nonsingular and $(AB)^{-1} = B^{-1}A^{-1}$.

$$\text{Proof} \Rightarrow (B^{-1}A^{-1})(AB) = B^{-1}(A^{-1}A)B = B^{-1}IB = B^{-1}B = I.$$

$$(AB)(B^{-1}A^{-1}) = A(BB^{-1})A^{-1} = AIA^{-1} = AA^{-1} = I.$$

$$\therefore (AB)^{-1} = B^{-1}A^{-1}.$$

□ If A is invertible, then A^T is invertible and $(A^T)^{-1} = (A^{-1})^T$.

$$\text{Proof} \Rightarrow A^T(A^T)^{-1} = A^T(A^{-1})^T$$

$$= (A^{-1}A)^T \Rightarrow I^T = I$$

$$(A^T)^{-1}A^T = (A^{-1})^T A^T$$

$$\Rightarrow (A^{-1}A)^T = I^T = I.$$

□ If A and B are invertible matrices, then $A+B$ is also invertible.

$$\text{Proof} \Rightarrow A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, B = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$\alpha = 1 \neq 0$
non singular.

$\alpha = -1 \neq 0$
non singular.

$$A+B = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \Rightarrow \alpha = 0 \text{ sing.}$$

⑤ The sum of singular matrices is also singular.

Proof \Rightarrow (False):-

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \text{ but } A+B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ non-singular.}$$

$\alpha=0$ $\alpha=0$ $\alpha=1$

⑥ $A^2 - B^2 = (A-B)(A+B)$, where A and B are matrices.

Proof \Rightarrow False

true in \mathbb{R} and \mathbb{C} only!

if A and B are commute: $AB = BA$, then:-

$$A^2 - B^2 = (A+B)(A-B)$$

⑦ $(A+B)^2 = A^2 + 2AB + B^2 \Rightarrow$ false.

⑧ If $AB = 0$, then $A = 0$ or $B = 0 \Rightarrow$ false.

Proof \Rightarrow $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \Rightarrow AB = 0$ but $A \neq 0$ and $B \neq 0$.

⑨ If $A^2 = 0$, then $A = 0 \Rightarrow$ false.

⑩ If $AB = AC$, then $B = C \Rightarrow$ false.

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$$

\rightarrow To be true \Rightarrow if A is invertible and $AB = AC \Rightarrow B = C$.

⑪ If $A^2 = A$, then $A = 0$ or $A = I \Rightarrow$ false.

⑫ If A is $n \times n$ matrix, such that $A^2 = A$, then $I + A$ is nonsingular and $(I + A)^{-1} = I - \frac{1}{2}A \Rightarrow$ True.

$$\begin{aligned} \text{Proof} \Rightarrow (I+A)(I-\frac{1}{2}A) &= I^2 - \frac{1}{2}IA + AI - \frac{1}{2}A^2 \\ &= I - \frac{1}{2}A + A - \frac{1}{2}A \quad \text{given.} \\ &= I + 0 = I. \end{aligned}$$

$$(I-\frac{1}{2}A)(I+A) = I.$$

$$\therefore (I+A)^{-1} = I - \frac{1}{2}A.$$

13] Let A be $n \times n$ matrix, then if $A^2 = 0$, then $I-A$ is nonsingular and $(I-A)^{-1} = I+A$. \Rightarrow True.

14] If A and B are $n \times n$ matrices. Then if $AB = A$ and $B \neq I$, then A must be singular. \Rightarrow True.

Proof \Rightarrow If A were nonsingular, then A^{-1} exists. Thus,

$$A^{-1}(AB), A^{-1}A \Rightarrow I B = I \Rightarrow B = I$$

which is a contradiction. Therefore, A must be singular.

1] If A is row equivalent to B , then B is row equivalent to A .

↳ True:-

Proof:- Suppose that A is row equivalent to B then

$A = (E_k E_{k-1} \dots E_1) B$, where E_1, E_2, \dots, E_k elementary matrices.

$$(E_k E_{k-1} \dots E_1)^{-1} A = \underbrace{(E_k \dots E_1)^{-1} (E_k \dots E_1)}_{I} B$$

$$\therefore B = E_1^{-1} E_2^{-1} \dots E_k^{-1} A$$

↳ B is row equivalent to A .

2] If A is row equivalent to B and B is row equivalent to C , then A is row equivalent to C .

↳ True:-

Proof:- We have:-

$$\text{and } A = (E_k E_{k-1} \dots E_1) B$$

$B = (F_k F_{k-1} \dots F_1) C$, where $E_1, \dots, E_k, F_1, \dots, F_k$ are elementary matrices then,

$$A = (E_k E_{k-1} \dots E_1 F_k F_{k-1} \dots F_1) C$$

↳ A is row equivalent to C .

3] Let A be an $n \times n$ matrix. Then the following statements are equivalent

a] A is nonsingular.

b] $Ax=0$ has only the trivial solution ($x=0$ the zero solution).

c] A is row equivalent to I_n .

Proof a] \rightarrow b] :- Suppose that A is nonsingular and y is a solution of $Ax=0$, i.e., $Ay=0$. Multiply both sides by A^{-1} from left, we get $A^{-1}(Ay) = A^{-1}0$. So $Iy=0$, i.e., $y=0$.

$b \Rightarrow c$:- Suppose that $Ax=0$ has only zero solution and suppose that A is not row equivalent to I , so the RREF of A has a free variable and so $Ax=0$ has infinitely many solutions which is a contradiction.

$c \Rightarrow a$:- Suppose that A is row equivalent to I_n , so there exist a finite sequence of elementary matrices E_1, E_2, \dots, E_k such that $(E_k E_{k-1} \dots E_1) I = A$. So, $A = E_k E_{k-1} \dots E_1$ and so $A^{-1} = E_1^{-1} E_2^{-1} \dots E_k^{-1}$.
 A is invertible (non singular).

4) If A is a 4×4 matrix and $a_1 + a_2 = a_3 + 2a_4$ then A must be singular. \Rightarrow True.
 $\hookrightarrow Ax=0$ has infinite.

Proof $\Rightarrow a_1 + a_2 - a_3 - 2a_4 = 0 \Rightarrow (1, 1, -1, -2)$ is a solution of $Ax=0$.

$\Rightarrow Ax=0$ has infinitely many solution.

$\Rightarrow A$ is singular. (Theorem)

5) If A has no LU -factorization, then A is non singular iff L is non singular.
 \hookrightarrow False.

Elementary matrices are all non singular

6) If A has an LU -factorization, then A is non singular iff U is non singular.
 \hookrightarrow True $\Rightarrow Ax=0 \Rightarrow L(Ux)=0 \Rightarrow Ux=L^{-1}0=0$.

7) If A has an LU -factorization, then A is row equivalent to I .
 \hookrightarrow True.

□ If A is an $n \times n$ matrix, then $\det(A^T) = \det(A) \Rightarrow |A^T| = |A|$.
↳ Proof: by induction.

① Lemma: Let A be an $n \times n$ matrix. then $a_{i1}A_{j1} + a_{i2}A_{j2} + \dots + a_{in}A_{jn}$

$$= \begin{cases} 0, & i \neq j \\ |A|, & i = j \end{cases}$$

Proof:- If $i \neq j$ then

$$a_{i1}A_{j1} + a_{i2}A_{j2} + \dots + a_{in}A_{jn} = a_{i1}A_{i1} + \dots + a_{in}A_{in} = |A|$$

if $i \neq j$, Let A^* be the matrix obtained from A by replacing the j^{th} row by the i^{th} row, i.e.,

$$A^* = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{i1} & a_{i2} & \dots & a_{in} \\ a_{j1} & a_{j2} & \dots & a_{jn} \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$

j^{th} row.
 i^{th} row.

since two rows of A^* are the same, so $|A^*| = 0$.

$$\begin{aligned} \text{So } 0, \det(A^*) &= a_{i1}A_{j1}^* + a_{i2}A_{j2}^* + \dots + a_{in}A_{jn}^* \\ &= a_{i1}A_{j1} + a_{i2}A_{j2} + \dots + a_{in}A_{jn} \end{aligned}$$

② Let E_1, \dots, E_k be elementary matrices then $|E_1 \dots E_k| = |E_1| \dots |E_k|$.
 Proof:- Use math. induction.

③ An $n \times n$ matrix A is singular iff $\det(A) = 0$.

Proof:-
 (i) \Rightarrow An $n \times n$ matrix A is non singular, iff $|A| \neq 0$.

(ii) \Rightarrow Let A be nonsingular. So, A is row equivalent to I_n .
 that is, there exist elementary matrices E_1, E_2, \dots, E_k such that
 $A \sim E_1 \dots E_k I_n$.

$$\text{So } |A| = |E_1| |E_2| \dots |E_k| \neq 0.$$

\Rightarrow conversely \Rightarrow suppose that $|A| \neq 0$ then the matrix A can be changed to RREF. With a finite number of row operations. That is, there exist elementary matrices E_1, \dots, E_k , and a matrix U in RREF such that $E_k E_{k-1} \dots E_1 U = A$. since

$|A| \neq 0$, so, $|U| \neq 0$, since, all E_i 's are invertible, and $|A|, |E_1|, |E_2|, \dots, |E_k|, |U|$
So, $|U| \neq 0$, and so A is invertible.

[4] If A, B are $n \times n$ matrices, then $\det(AB) = \det(A) \det(B)$.

→ proof:- If A is singular then $|A| = 0$ and so AB is singular
and therefore, $|AB| = 0 = |A| |B|$

If A is nonsingular, then A is row equivalent to I_n . That is,
 $A \rightarrow E_1 \dots E_k I_n = E_1 \dots E_k$.

$$\text{Thus, } |AB| = |E_1 E_2 \dots E_k B| \\ = |E_1| |E_2| \dots |E_k| |B| = |A| |B|.$$

[5] $|A+B| = |A| + |B|$, where $+$ is defined \Rightarrow False.

[6] $|A^n| = |A|^n$, $n \geq 1$, $\dots \Rightarrow$ True.

[3] $|kA| = k^n |A|$ where $A_{n \times n}$, $k \in \mathbb{R}$. \Rightarrow True.

[4] If A is nonsingular, then $\det(A^{-1}) = \frac{1}{\det A} \Rightarrow$ True.

→ proof:- A nonsingular $|A^{-1}| = \frac{1}{|A|}$

$$A^{-1} A = I$$

$$|A^{-1} A| = |I|$$

$$|A^{-1}| |A| = 1 \Rightarrow |A^{-1}| = \frac{1}{|A|}.$$

5 If $A^2 = A$, then $|A| = 0$ or $|A| = 1 \Rightarrow$ false.

↳ proof $\Rightarrow A^2 = A \Rightarrow |A| = 0$ or $|A| = 1$.

$$|A^2| = |A|$$

$$|A|^2 = |A|$$

$$|A|(|A| - 1) = 0$$

$$|A| = 0 \text{ or } |A| = 1$$

6 If $A^T A = I$, then $|A| = \pm 1 \Rightarrow$ true.

↳ proof: if $|A^T A| = |I|$

$$|A^T| |A| = 1$$

$$|A| |A| = 1 \Rightarrow |A|^2 = 1 \Rightarrow |A| = \pm 1.$$

7 If $A_{n \times n}$ is skew symmetric and n is odd, then A must be singular.

↳ True. \Rightarrow proof $\Rightarrow A^T = -A$.

$$|A^T| = |-A|$$

$$|A| = (-1)^n |A| = -|A|, \text{ odd}$$

$$\Rightarrow 2|A| = 0$$

$$|A| = 0 \Rightarrow A \text{ singular.}$$

8 If $A_{n \times n}$ is skew symmetric and n is even, then A must be nonsingular.

↳ False. \Rightarrow proof \Rightarrow same as 7 but n is even.

9 Let $A_{n \times n}, B_{n \times n}$. then AB is nonsingular iff A and B are both nonsingular.

↳ True. \Rightarrow proof $\Rightarrow AB$ is nonsingular $\Leftrightarrow |AB| \neq 0$

$$|A| |B| \neq 0.$$

$$\Leftrightarrow |A| \neq 0 \text{ and } |B| \neq 0$$

So A and B both nonsingular.

10 If A, B , and C are 3×3 matrices, $|A|=9$, $|B|=2$, $|C|=3$. then, $|4 C^T B A^{-1}| = \frac{128}{3} \Rightarrow \text{True}$.

→ proof $|4 C^T B A^{-1}| = 4^3 |C^T| |B| |A^{-1}|$
 $= 64 |C| |B| \frac{1}{|A|} = 64 \times 3 \times 2 \times \frac{1}{9}$
 $= \frac{128}{3}$

11 Let A be an $m \times n$ matrix. Explain why the matrix multiplications $A^T A$ and $A A^T$ are possible.

→ A is $m \times n$ matrix and A^T is $n \times m$ matrix.

$A^T A$ is $n \times m$ $m \times n$ matrix which is valid.

$A A^T$ is $m \times n$ $n \times m$ matrix which is valid.

12 A matrix A said to be skew symmetric if $A^T = -A$. show that if a matrix is skew symmetric, then its diagonal entries must all be 0.

→ A is skew symmetric matrix, so $A^T = -A$ which is a square matrix a_{ij} are the matrices on A and b_{ij} the entries of A^T . So we have $b_{ij} = a_{ji}$ since $A^T = -A$ so $b_{ij} = -a_{ji}$. Thus $a_{ji} = -a_{ij}$, and if we take $i=j$ we have $a_{ij} = -a_{ij}$ which gives us that $a_{ij}=0$, which are the diagonal entries.

① $A \operatorname{adj}(A) = |A| I_n$.

→ Proof: The i^{th} entry of $A \operatorname{adj}(A)$ is:-

$$A \operatorname{adj}(A)_{ii} = a_{i1} A_{j1} + \dots + a_{in} A_{jn} = \begin{cases} |A|, & i=j \\ 0, & i \neq j \end{cases} = |A| I_n.$$

② $A^{-1} = \frac{1}{|A|} \operatorname{adj}(A)$.

→ Proof: Since A is nonsingular, then A^{-1} exists. From last theorem we know, $A \operatorname{adj}(A) = |A| I_n$. Multiply both sides by A^{-1} from left:-

$$A^{-1} A \operatorname{adj}(A) = |A| A^{-1} I_n = |A| A^{-1}.$$

$$\operatorname{adj}(A) = |A| A^{-1} \text{ and so, since } |A| \neq 0, A^{-1} = \frac{1}{|A|} \operatorname{adj}(A).$$

③ $x_i = \frac{|A_i|}{|A|}$, $i=1, 2, \dots, n$.

→ Proof: since $x = A^{-1} b = \frac{1}{|A|} \operatorname{adj}(A) b$.

$$\text{it follows that } \Rightarrow x_i = \frac{b_1 A_{1i} + b_2 A_{2i} + \dots + b_n A_{ni}}{|A|} = \frac{|A_i|}{|A|}.$$

④ Let A be a nonsingular $n \times n$ matrix with $n \geq 1$. Show that $|\operatorname{adj} A| = |A|^{n-1}$.

→ If A is nonsingular, then $|A| \neq 0$ and hence $\operatorname{adj} A = |A| A^{-1}$ is also nonsingular $\Rightarrow |\operatorname{adj} A| = |(|A| A^{-1})| = |A|^n |A^{-1}| = |A|^n \frac{1}{|A|} = |A|^{n-1}$.

constant

⑤ Show that if A is nonsingular, then $\operatorname{adj} A$ is nonsingular and $(\operatorname{adj} A)^{-1} = |A|^{-1} A = \operatorname{adj} A^{-1}$.

→ If A nonsingular, then $|A| \neq 0$ and hence $\operatorname{adj} A = |A| A^{-1}$ is also nonsingular.

To prove the equation, we have $\Rightarrow (\operatorname{adj} A)^{-1} = (|A| A^{-1})^{-1} = \frac{1}{|A|} (A^{-1})^{-1} = \frac{1}{|A|} A = |A|^{-1} A \Rightarrow$ ①
and $\operatorname{adj} A^{-1} = |A^{-1}| (A^{-1})^{-1} = |A|^{-1} A$, hence ① and ② gives the equation.

6] if $|\text{adj } A| = |A|$ and non singular, then A is 2×2 matrix \Rightarrow false.

$$\rightarrow I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow |\text{adj}(I)| = |I|^3 = (1)^3 = 1$$

$$|I| = 1$$

$$\rightarrow |\text{adj } I| = |I| \text{ but, } I \text{ is } 3 \times 3.$$

7] Let A be $n \times n$ matrix (nonsingular), then $\text{adj}(\text{adj } A) = |A|^{n-2} A$

Proof:- since A is nonsingular, then $|A| \neq 0$ and hence $\text{adj } A = |A| A^{-1}$

$$\Rightarrow \text{adj}(\text{adj } A) \cdot \text{adj}(|A| A^{-1}) = |A| |A^{-1}|^n (|A| A^{-1})^{-1} = |A|^n |A^{-1}| \frac{1}{|A|} (A^{-1})^{-1}$$

$$= |A|^n \frac{1}{|A|} \cdot \frac{|A|}{|A|} \cdot A = |A|^{n-2} A.$$

8] Show that if $|A| = 1$, then $\text{adj}(\text{adj } A) = A$.

\rightarrow This question is special case of the above example. We proved that $\text{adj}(\text{adj } A) = |A|^{n-2} A$.

if $|A| = 1$, then $\text{adj}(\text{adj } A) = (1)^{n-2} A = A$.

9] Let A and B are $n \times n$ matrices. show that if $AB = A$ and $B \neq I$ then A must be singular.

$\rightarrow AB = A$, $B \neq I$ suppose A is nonsingular and A^{-1} exists such that $A^{-1}A = AA^{-1} = I$ multiply by A^{-1} .

$$A^{-1}(AB) = (A^{-1}A)B = A^{-1}A \Rightarrow B = I \text{ (contradiction).}$$

Therefore A is singular.

10] Prove that if A is nonsingular then A^T is nonsingular and $(A^T)^{-1} = (A^{-1})^T$

$$\rightarrow A^{-1}A = I \Rightarrow (A^{-1}A)^T = I^T$$

$$= A^T (A^{-1})^T = I^T$$

$$(A^T)^{-1} A^T (A^{-1})^T = (A^T)^{-1} I^T$$

$$(A^{-1})^T = (A^T)^{-1}$$

1) $\mathbb{Q} = \{x = \frac{a}{b} \text{ are integers } b \neq 0\}$ is not a vector space.

$$\rightarrow x = \frac{2}{5} \in \mathbb{Q}, \alpha = \sqrt{2}$$

$$\alpha x = \sqrt{2} \times \frac{2}{5} = \frac{2\sqrt{2}}{5} \notin \mathbb{Q}.$$

2) $\mathbb{Z} = \{0, \pm 1, \pm 2, \dots\}$ is not a vector space.

$$\rightarrow \alpha = \frac{1}{2}, x = 5 \in \mathbb{Z} \text{ but}$$

$$\alpha x = \frac{1}{2} \times 5 = \frac{5}{2} \notin \mathbb{Z}.$$

3) Irrationals, $(\sqrt{2}, \sqrt{3}, e, \pi, \dots)$ is not a vector space since:

$$\rightarrow \alpha = \frac{1}{2}, x = \sqrt{2} \in \mathbb{Q}.$$

$$\text{but } \alpha x = \sqrt{2} \cdot \sqrt{2} = 2 \notin \mathbb{Q}.$$

4) $0\vec{v} = \vec{0}$

$$\rightarrow \text{proof: } 0 = 0 + 0, \text{ so } (0+0)\vec{v} = 0\vec{v}.$$

thus $0\vec{v} + 0\vec{v} = 0\vec{v}$ add to both sides

$$-0\vec{v}. \text{ So, } 0\vec{v} + 0\vec{v} + -0\vec{v} = 0\vec{v} + -0\vec{v}.$$

$$\Rightarrow 0\vec{v} + \vec{0} = \vec{0}. \text{ Hence } 0\vec{v} = \vec{0}.$$

5) If $x+y = \vec{0}$, then $y = -x$.

$$\rightarrow \text{proof: Add } -x \text{ to both sides of } x+y = \vec{0}.$$

$$\text{So, } -x + x + y = -x + \vec{0}. \text{ Thus, } \vec{0} + y = -x + \vec{0}.$$

$$\Rightarrow y = -x.$$

6) $-1 \cdot \vec{v} = -\vec{v}, \forall \vec{v} \in V$.

$$\rightarrow 0 = 1 + -1, \text{ so } (1 + -1)\vec{v} = 0\vec{v} = \vec{0}.$$

$$\text{Thus, } 1\vec{v} + -1\vec{v} = \vec{0}, \text{ so } \vec{v} + -1\vec{v} = \vec{0}.$$

$$\Rightarrow -\vec{v} + \vec{v} + -1\vec{v} = -\vec{v} + \vec{0} = -\vec{v}.$$

$$\Rightarrow 0 + -1\vec{v} = -\vec{v} \Rightarrow \text{Thus } -1\vec{v} = -\vec{v}.$$

1] Let S be a subspace of a vector space V . Then $\vec{0} \in S$.

↳ proof \Rightarrow since S is a subspace of V , then $S \neq \emptyset$. Let $x \in S$, so $0x = \vec{0} \in S$.

2] $S \cap T$.

↳ since $0 \in S$, $0 \in T$, then $0 \in S \cap T \Rightarrow S \cap T \neq \emptyset$.

* Let $x, y \in S \cap T$, then $x, y \in S$ and $x, y \in T$

$\Rightarrow x+y \in S$ and $x+y \in T$.

$x+y \in S \cap T$.

* Let $\alpha \in \mathbb{R}$ and $x \in S \cap T$.

since $x \in S \cap T$, then $x \in S$ and $x \in T$. since S and T are subspaces, it follows $\alpha x \in S$ and $\alpha x \in T$. $\Rightarrow \alpha x \in S \cap T$.

3] $S \cup T$ is not always a subspace of V .

↳ Let $S = \{(x, 0) : x \in \mathbb{R}\}$.

$T = \{(0, y) : y \in \mathbb{R}\}$.

Notice that S and T are subspaces of \mathbb{R}^2 but $S \cup T = \{(x, y) : x \text{ or } y \text{ is zero}\}$ is not subspace, for example, $(0, 1), (1, 0) \in S \cup T$, but $(0, 1) + (1, 0) = (1, 1) \notin S \cup T$.

4] Let A be $m \times n$ matrix, then $N(A)$ is a subspace of \mathbb{R}^n .

↳ [i] since $A\vec{0} = \vec{0}$, then $\vec{0} \in N(A)$

$\therefore N(A) \neq \emptyset$

[ii] Let $x, y \in N(A)$. Then $Ax = \vec{0}$ and $Ay = \vec{0}$, so $A(x+y) = Ax + Ay$
 $= \vec{0} + \vec{0} = \vec{0}$.

$\Rightarrow x+y \in N(A)$.

[iii] Let $x \in N(A)$, $\alpha \in \mathbb{R}$. Then $Ax = \vec{0}$, so $A(\alpha x) = \alpha Ax$
 $= \alpha \vec{0} = \vec{0}$.

$\alpha x \in N(A)$.

5] Let V be a vector space and let $v_1, v_2, \dots, v_k \in V$. then $\text{span}(v_1, \dots, v_k)$ is a subspace of V .

[i] since $\vec{0} = 0 \cdot v_1 + 0 \cdot v_2 + \dots + 0 \cdot v_k$, then $\vec{0} \in \text{span}(v_1, \dots, v_k)$. That is, $\text{span}(v_1, \dots, v_k) \neq \emptyset$.

[ii] Let $x, y \in \text{span}(v_1, \dots, v_k)$. Then $x = \alpha_1 v_1 + \dots + \alpha_k v_k$.

$$y = \beta_1 v_1 + \dots + \beta_k v_k.$$

$$\text{So, } x + y = (\alpha_1 + \beta_1) v_1 + \dots + (\alpha_k + \beta_k) v_k.$$

$$= \gamma_1 v_1 + \dots + \gamma_k v_k$$

$$\Rightarrow x + y \in \text{span}(v_1, \dots, v_k).$$

[iii] Let $x \in \text{span}(v_1, \dots, v_k)$, $\alpha \in \mathbb{R}$.

$$\text{Then } \alpha x = \alpha (c_1 v_1 + c_2 v_2 + \dots + c_n v_n)$$

$$= (\alpha c_1) v_1 + (\alpha c_2) v_2 + \dots + (\alpha c_n) v_n.$$

$$\Rightarrow \alpha x \in \text{span}(v_1, \dots, v_k).$$

therefore, $\text{span}(v_1, \dots, v_k)$ is subspace of V .

6] Let A be an $n \times n$ matrix. show that if $A^2 = 0$, then $I - A$ is nonsingular and $(I - A)^{-1} = I + A$.

$$(I + A)(I - A) = I + A - A - A^2 = I$$

$$(I - A)(I + A) = I - A + A - A^2 = I \text{ therefore } I - A \text{ is non singular and } (I - A)^{-1} = I + A.$$

Let A be an $n \times n$ matrix. Then the following statements are equivalent:

A- A is non singular.

B- $Ax = 0$ has only the trivial solution ($x = 0$ The zero solution).

C- A is row equivalent to I_n .

أذا كان A غير قابل للعكس،
فإن $Ax = 0$ له حلول غير تافهة.

* Let A be $n \times n$ matrix and α is scalar. show that $|\alpha A| = \alpha^n |A|$.

$\rightarrow \alpha I$ is a diagonal matrix with all diagonal entries equal α . So $|\alpha I| = \alpha^n$.

$$\alpha A = \alpha (I A) \Rightarrow (\alpha I) A \Rightarrow |\alpha A| = |(\alpha I) A| = |\alpha I| |A| = \alpha^n |A|.$$