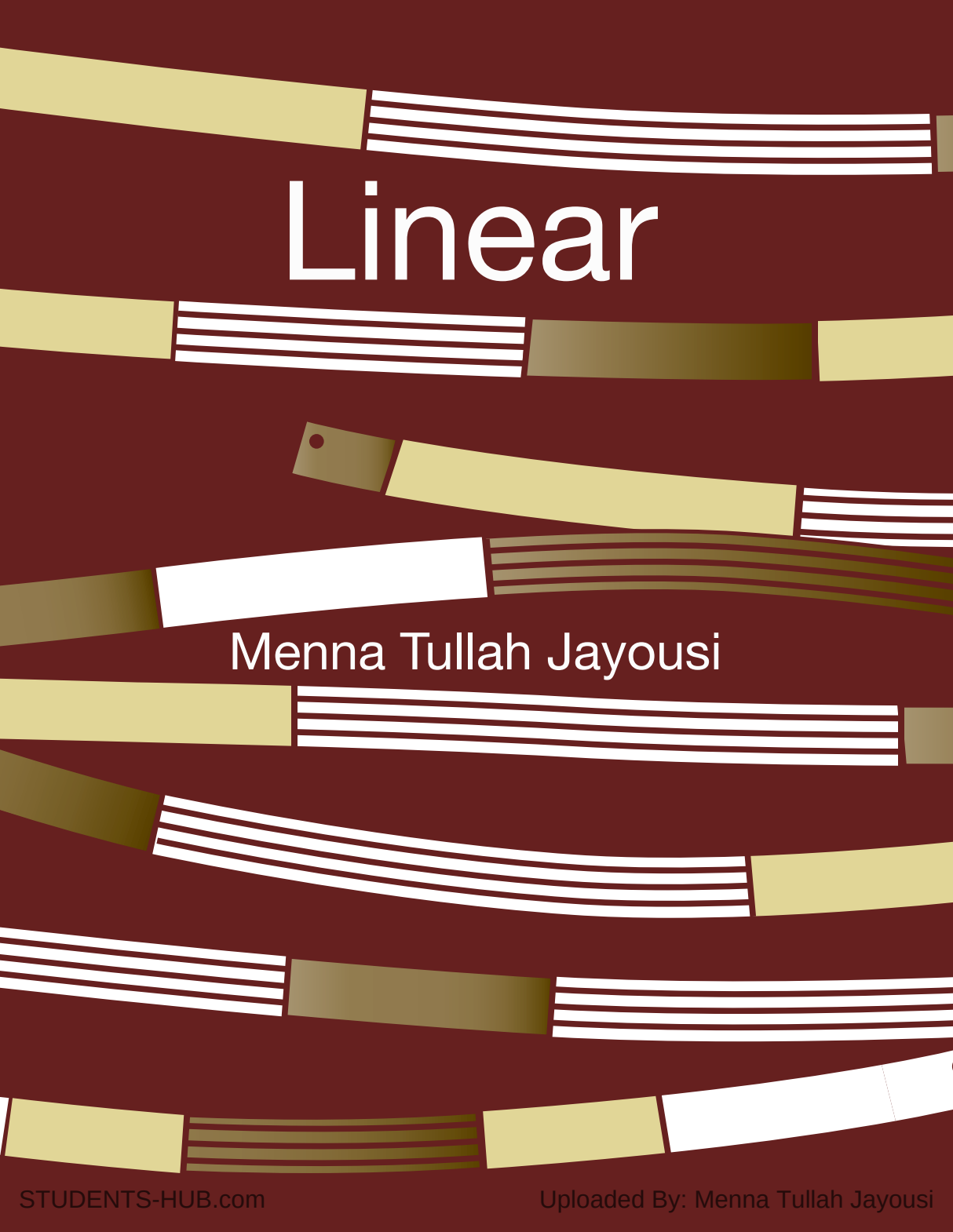




Linear

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Chapter one :

1.1: system of linear equations:

A linear equation : $a_1x_1+a_2x_2+\dots+a_nx_n=b$, where x are variables, a, b are real numbers.

$$2x_1-x_2+x_3=5$$

$$x_1+3x_2+x_3=6$$

Is 2×3 linear system.

$m \times n \Rightarrow$ number of unknowns (variables).
↓
number of equations.

A non linear system : at least one of the equations in the system are nonlinear :

Exp $\Rightarrow x+y=1$

$$x+y=5$$

Remark : in general there are three possibilities for 2×2 linear system:

- 1- the lines intersect at a point (unique solution).
- 2- they are parallel (no solution).
- 3- both equations represent the same line (infinite solution).

Remark : in general there are three possibilities for $m \times n$ linear system:

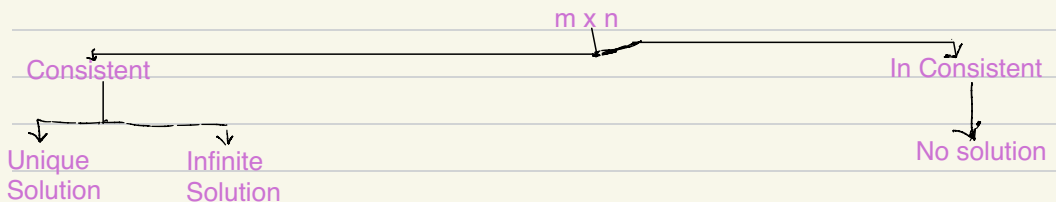
- 1- (unique solution).
- 2- (no solution).
- 3- (infinite solution).

Ex: $\begin{cases} 4x_1 + 2x_2 = 6 \\ -4x_1 + 2x_2 = -6 \end{cases}$

$0 = 0$

free $\begin{cases} x_1 = 2x_2 - 3 \\ x_2 = 2x_2 - 3 \end{cases} \rightarrow$ leading $\begin{cases} x_2 = 2x_2 - 3 \end{cases}$

The solution set = $\{(x_1, x_2) = (t, 2t-3) \mid t \in \mathbb{R}\}$



Equivalent systems :

Two systems of equations are called equivalent systems if they have the same variables (unknowns) and the same solution set.

Square system: If $m=n$ and it is called an $n \times n$ linear system.

Strict triangular form :

1- $n \times n$ system .

2- if in the k equation the coefficient of the first $k-1$ variables are all zero and the coefficient of x_k is nonzero ($k=1,2,3,\dots,n$)

Exp:

$$x_1 + x_2 + x_3 = 8$$

$$0 + 2x_2 + x_3 = 5$$

$$0 + 0 + x_3 = 9$$

This example is strict triangular form.

1- the system is strict triangular form is easy to solve and has a unique solution.

2- we have the system by backward substitution method. برجع رجوع بالحل.

How to transform a system in strict triangular form?

1- interchange two rows

Augmented matrix = $[A:b]$

2- multiply a row by a nonzero constant

3- replace a row by its sum with a multiple of another row.

Coefficient matrix = $[A]$ بكتب معاملات الاكسبات كمصفوفة

7. The two systems

$$2x_1 + x_2 = 3$$

and

$$2x_1 + x_2 = -1$$

$$4x_1 + 3x_2 = 5$$

$$4x_1 + 3x_2 = 1$$

have the same coefficient matrix but different right-hand sides. Solve both systems simultaneously by eliminating the first entry in the second row of the augmented matrix

$$\left[\begin{array}{cc|cc} 2 & 1 & 3 & -1 \\ 4 & 3 & 5 & 1 \end{array} \right]$$

and then performing back substitutions for each of the columns corresponding to the right-hand sides.

$$\left[\begin{array}{cc|cc} 2 & 1 & 3 & -1 \\ 0 & 1 & -1 & 3 \end{array} \right]$$

دائرة هار باس لهار :-

$$\left[\begin{array}{cc|cc} 2 & 1 & 3 & -1 \\ 0 & 1 & -1 & 3 \end{array} \right]$$

$x_2 = -1$

$2x_1 - 1 = 3 \Rightarrow x_1 = 2$

$x_2 = 3$

$2x_1 + 3 = -1 \Rightarrow x_1 = -2$

1.2: Row Echelon Form(REF)

1- the first nonzero entry in each nonzero row is 1 called the leading one or the pivot 1.

2- the leading 1 in the k row is to the right of the leading 1 in the (k-1) row.

3- zero rows are below the nonzero rows.

$$\text{Exp} = \begin{bmatrix} 1 & 4 & 2 \\ 0 & 1 & 5 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow \text{REF.}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \text{ is not REF.}$$

$$\begin{bmatrix} 2 & 4 & 6 \\ 0 & 3 & 5 \\ 0 & 0 & 4 \end{bmatrix} \Rightarrow \text{is not REF.}$$

$$\begin{bmatrix} 1 & 3 & 1 & 0 \\ 0 & 0 & 1 & 3 \\ 0 & 0 & 0 & 0 \end{bmatrix} \Rightarrow \text{is REF.}$$

Gaussian elimination method (REF)

Gaussian Jordan elimination method (RREF)

1- the matrix is in REF.

2- the first nonzero entry in each row 1's is the only nonzero entry in its column.

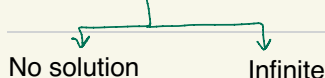
$$A = \begin{bmatrix} 1 & 0 & 0 & 3 \\ 0 & 1 & 0 & 2 \\ 0 & 0 & 1 & 1 \end{bmatrix} \text{ is RREF.}$$

$$B = \begin{bmatrix} 0 & 1 & 2 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \text{ is RREF.}$$

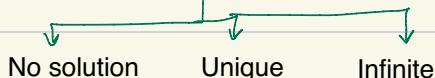
$$C = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \text{ is RREF.}$$

$$D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \text{ is Not RREF.}$$

Under determined : $m < n$



Over determined : $m > n$



Homogeneous system : $Ax=0$

Homogeneous system is always consistent since there is a solution called the zero solution or the trivial solution

Under determined homogeneous linear system has always infinite solution

$$\left[\begin{array}{ccc|c} 0 & 1 & 0 & 2 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{array} \right] \Rightarrow \begin{cases} x_1 = t \\ x_2 = 2 \\ x_3 = -1 \end{cases} \Rightarrow (t, 2, -1) \text{ الجواب}$$

$$\left[\begin{array}{ccc|c} 1 & 2 & 1 & 0 \\ 2 & 5 & 3 & 0 \\ 1 & 1 & 1 & 0 \end{array} \right] \Rightarrow \text{inconsistent system} \quad \text{REF} \quad \Rightarrow \text{inconsistent system} \quad \text{homogeneous.}$$

example $\Rightarrow$ Let $Ax=b$ be a linear system whose augmented matrix $(A|b)$ has reduced row echelon form:

$$\left[\begin{array}{ccccc|c} 1 & 2 & 0 & 3 & 1 & -2 \\ 0 & 0 & 1 & 2 & 4 & 5 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right], \text{ if } a_1 = \begin{bmatrix} 1 \\ 1 \\ 3 \\ 4 \end{bmatrix} \text{ and } a_3 = \begin{bmatrix} 2 \\ -1 \\ 1 \\ 3 \end{bmatrix} \text{ determine } b.$$

*if A and B are singular then $A+B$ is not also non singular.

1.3 : Matrix arithmetic

$m \times n$ is the size (order) or the dimension of $A_{m \times n}$

For similarity we use the notation : $A = a_{ij} : i=1,2,\dots,m$
 $J=1,2,\dots,n$

a_{45}
الصف ←
العمود →

Column vector is an $m \times 1$ matrix. $\Rightarrow A = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$

Row vector is an $1 \times n$ matrix. $B = [1. 3. 4. 7.]$

عدد الصفوف : 2
عدد الأعمدة : 3
 $R : \begin{bmatrix} 10 & 5 & 7 \\ 2 & 13 & 6 \end{bmatrix}$

* column vectors $\Rightarrow a_j = \begin{pmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{mj} \end{pmatrix}$

* Row vector $\Rightarrow A = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_m \end{pmatrix}$

The equality matrices : two matrices A and B are equal iff they have the same size and $a_{ij}=b_{ij}$.

Properties of addition and scalar multiplication:

1- $k(A+B)=kA+kB$.

2- $(k r)A = r(kA)=k(rA)$.

3- $A+B=B+A$.

4- $A+(B+C)=(A+B)+C$.

5- $A+0=0+A=A$.

6- $A-A=A+(-A)=0$.

$-A$ is called the additive inverse of A.

When I want to multiply two matrices A and B they should have the same size.

$\hookrightarrow A = \begin{bmatrix} 1 & 3 \\ 6 & -1 \end{bmatrix} 2 \times 2$ $B = \begin{bmatrix} 1 & 5 & 2 \\ 0 & 1 & -3 \end{bmatrix} 2 \times 3$

→ explain
← by the exp

$\hookrightarrow AB = \begin{bmatrix} 1 \times 1 + 3 \times 0 & 1 \times 5 + 3 \times 1 & 1 \times 2 + 3 \times (-3) \\ 6 \times 1 + (-1) \times 0 & 6 \times 5 + (-1) \times 1 & 6 \times 2 + (-1) \times (-3) \end{bmatrix} 2 \times 3$
الصفوف :
الأعمدة : نتيج

$AB \neq BA$

If it is told me to write the system in a matrix form :

$$4x_1 + 2x_2 + x_3 = 1.$$

$$5x_1 + 3x_2 + 7x_3 = 2.$$

$$\begin{bmatrix} 4 & 2 & 1 \\ 5 & 3 & 7 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

A x b

write as

$$x_1 \begin{bmatrix} 4 \\ 5 \end{bmatrix} + x_2 \begin{bmatrix} 2 \\ 3 \end{bmatrix} + x_3 \begin{bmatrix} 1 \\ 7 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}.$$

Linear combination : $c_1a_1 + c_2a_2 + \dots + c_n a_n$

↳ exp $\Rightarrow I \Rightarrow b = \begin{pmatrix} 2 \\ 24 \end{pmatrix}$ a linear combination of $a_1 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$, $a_2 = \begin{pmatrix} 0 \\ 5 \end{pmatrix}$?

Answer $\Rightarrow b = c_1 a_1 + c_2 a_2$

$$\begin{pmatrix} 2 \\ 24 \end{pmatrix} = c_1 \begin{pmatrix} 1 \\ 2 \end{pmatrix} + c_2 \begin{pmatrix} 0 \\ 5 \end{pmatrix} \Rightarrow \quad 2 = c_1 + \cancel{c_2 \cdot 0} \Rightarrow \boxed{c_1 = 2}$$

$$24 = 2c_1 + 5c_2$$

$$24 = 4 + 5c_2 \Rightarrow \boxed{c_2 = 4}$$

$b = 2a_1 + 4a_2 \Rightarrow b$ is a linear combination of a_1 and a_2 .

- not linear combination if c_1 and c_2 are not integers

The transpose of a matrix : $\Rightarrow A^T = (a_{ij})^T = (a_{ji}) = B \text{ } n \times m$

↳ exp $\Rightarrow A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \Rightarrow A^T = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$

An $n \times n$ matrix A is said to be symmetric if $A^T = A$

An $n \times n$ matrix A is said to be **skew** symmetric if $A^T = -A$

The zero matrix is always skew or symmetric . $\Rightarrow A^T = A \quad \text{or} \quad -A^T = A \quad \Rightarrow \quad 0 = 2A \Rightarrow \boxed{A = 0}$

Properties of the transpose :

1- $(A^T)^T = A$.

2- $(A \mp B)^T = A^T \mp B^T$.

3- $(KA)^T = KA^T$.

4- $(AB)^T = B^T A^T$.

5- If A, B are symmetric, then A+B is symmetric. $\Rightarrow (A+B)^T = A^T + B^T = A + B$.

6- If A is symmetric, then kA is symmetric. $\Rightarrow (\alpha A)^T = \alpha A^T = \alpha A$ is symmetric.

1.4: Matrix algebra:

$$A^n = \begin{bmatrix} 2^{n-1} & 2^{n-1} \\ 2^{n-1} & 2^{n-1} \end{bmatrix}$$

The identity matrix : $\Rightarrow I = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$

example $\Rightarrow I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$.

$n \times n$ matrix must be square

$$IB=BI=B$$

Matrix inversion :

An $n \times n$ matrix A is said to be invertible if there exists a matrix B such that $AB=BA=I$, the matrix B is called the inverse or multiplicative inverse of A denoted by A^{-1} . If A^{-1} does not exist, then A has no inverse (A is called singular or invertible).

$$A^{-1} \neq \frac{1}{A}$$

How to find the inverse of 2×2 ?

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \Rightarrow \alpha = ad - bc \neq 0 \text{ then } A^{-1} = \frac{1}{\alpha} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

If $\alpha = 0$, the A is singular.

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

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

Algebraic rules for inverses:

1- The inverse of A if exists is unique.

$$2- (A^{-1})^{-1} = A.$$

$$3- (kA)^{-1} = 1/k (A^{-1})$$

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

$$5- [(AB)^T]^{-1} = (A^T)^{-1} (B^T)^{-1}$$

1.5: elementary matrices :

A matrix E is called an elementary matrix if it is obtained from the identity I_n by performing exactly one row operations.

There are three types :

- 1- interchanging any two rows of I_n .
- 2- multiplying any row of I_n by a nonzero constant.
- 3- adding a multiple of one row of I_n to another row of I_n .

If E is an elementary matrix, then E is nonsingular (invertible) and E^{-1} is an elementary of the same type .

A matrix B is row equivalent to a matrix A : $B = E_k E_{k-1} \dots E_1 A$.

Example $\Rightarrow$ if we have $A = \begin{bmatrix} 1 & 2 & 4 \\ 2 & 1 & 3 \\ 1 & 0 & 2 \end{bmatrix}$, $B = \begin{bmatrix} 1 & 2 & 4 \\ 2 & 1 & 3 \\ 2 & 2 & 6 \end{bmatrix}$

find an elementary matrix E such that $EA=B \Rightarrow B$ is row equivalent to A .

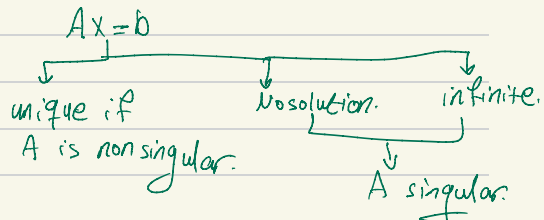
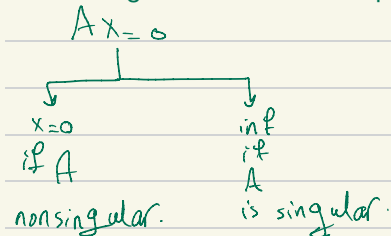
أول صفين في A and B هي نفسها. هسة بدى أوجد E اللي هتغير A بـ B .
أول صف B هسة بلاصته زي الأول، والثاني في A نغير نفس الطولاني
اللي في B $\Leftarrow E = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \Leftarrow$ جمع الصف الأول، والثاني $\Leftarrow E = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$
هذه بتكون الجواب.

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

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

The system $Ax=b$, where A is $n \times n$, has a unique solution, iff A is nonsingular (The solution is $x=A^{-1}b$).

If A is nonsingular then A is row equivalent to I .



To find the inverse for 3x3 matrix and above :

→ I will take an example $\Rightarrow A = \begin{bmatrix} 1 & 1 & 2 \\ 0 & 1 & 2 \\ 2 & 0 & 1 \end{bmatrix}$, find A^{-1} .

$[A | I_3] = \left[\begin{array}{ccc|ccc} 1 & 1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 2 & 0 & 1 & 0 \\ 2 & 0 & 1 & 0 & 0 & 1 \end{array} \right]$

RAFF
 الجداول بـ I_3 عن طريق
 والنتيجة هتكون A^{-1} هو الـ

(A is singular) inverse لها

If he tells me to solve the system $Ax=b$ using the inverse of A :

Example $\Rightarrow \begin{cases} x_1 + x_2 + 2x_3 = -2 \\ x_2 + 2x_3 = 3 \\ 2x_1 + x_3 = 0 \end{cases} \Rightarrow \begin{bmatrix} 1 & 1 & 2 \\ 0 & 1 & 2 \\ 2 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -2 \\ 3 \\ 0 \end{bmatrix}$

$A \quad x \quad b$

$x = A^{-1} b$

بـ A^{-1} و b عن طريق
 الطريقة المباشرة
 ضرب الـ A^{-1}

Diagonal and Triangular matrices:

A is said to be triangular if it is either upper or lower triangular.

$A = \begin{bmatrix} 3 & 2 & 1 \\ 0 & 2 & 1 \\ 0 & 0 & 5 \end{bmatrix}$ upper triangular.

$B = \begin{bmatrix} 1 & 0 & 0 \\ 6 & 0 & 0 \\ 0 & 4 & 3 \end{bmatrix}$ lower triangular.

$C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 4 \end{bmatrix}$ Diagonal.

$D = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ diagonal and triangular.

Every Diagonal is triangular.

LU factorization : (Triangular factorization)

$$\hookrightarrow A = LU.$$

*not every matrix has an LU factorization.

□ كيف بوجد الـ L ؟ خطية الـ A $\Rightarrow$ $\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ خطية ههههه.

□ في طريقتين بوجد فاهم الـ L في الطريقة الاولى المعادلات التي استعملناها مثانه الفهم في الـ L بوجد وبشكل خاصاهم وظاههم هونه.

$$L = \begin{bmatrix} 1 & 0 & 0 \\ L_{21} & 1 & 0 \\ L_{31} & L_{32} & 1 \end{bmatrix}$$

الطريقة الاصحى والاطول $\Rightarrow$ بعمل الـ E_1 و الـ E_2 و الـ E_3 خطية اول معادلات $\Rightarrow$ $E_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ استعملنا الـ L خطية هونه. ونفس الامر لـ E_2 و E_3 .

$$E_3 E_2 E_1 A = U \quad \text{بعدى}$$

$$A = (E_3 E_2 E_1)^{-1} U.$$

$$A = E_1^{-1} E_2^{-1} E_3^{-1} U.$$

$$[E_1 | I] \quad \text{وههههنا بوجدهم.}$$

1- if A has an LU factorization , then A is nonsingular iff L is nonsingular. False because L is always nonsingular.

2- if A has an LU factorization , then A is nonsingular iff U is nonsingular. True.

3- if A has an LU factorization , then A is row equivalent to U. True.

2.1: The determinant of a matrix

If A was 2x2 matrix then the $\det(A)$ or the $|A|$ is =

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \Rightarrow \begin{vmatrix} a & b \\ c & d \end{vmatrix} = \text{ad} - \text{bc}$$

2x2 matrix

Then an nxn matrix A is nonsingular iff $|A| \neq 0$.

$$|AB| = |BA| = |B||A| = |A||B|$$

Cofactor method :

$\rightarrow m_{ij} = \text{the minor of } a_{ij} = |M_{ij}|$

$$A_{ij} = (-1)^{i+j} |M_{ij}|$$

example $\Rightarrow$ if $A = \begin{bmatrix} 2 & 5 & 4 \\ 3 & 1 & 2 \\ 5 & 4 & 6 \end{bmatrix}$ then $A_{32} = (-1)^{3+2} m_{32} \rightarrow$ $= -1 \begin{vmatrix} 2 & 4 \\ 3 & 2 \end{vmatrix} = -8$

2x2 matrix

If A was nxn matrix then the $\det(A)$ or the $|A|$ is = $\Rightarrow$ 2x2 matrix

$$\rightarrow a_{11}A_{11} + a_{12}A_{12} + \dots + a_{1n}A_{1n}, \text{ if } n \geq 2$$

example $\Rightarrow$ if $A = \begin{bmatrix} 3 & 2 & 4 \\ 1 & -2 & 3 \\ 2 & 3 & 2 \end{bmatrix}$ find $|A|$

$$|A| = a_{11}A_{11} + a_{12}A_{12} + a_{13}A_{13}$$

$$= 3(-1)^1 m_{11} + 2(-1)^2 m_{12} + 4(-1)^3 m_{13}$$

$$= -3$$

if we have $\begin{bmatrix} 6 & 0 & 0 \\ 8 & 2 & 0 \\ 1 & 5 & 6 \end{bmatrix} = 6 \times 2 \times 6 = 6 \times 12 = 72$

$\Rightarrow$ triangular matrix $\Rightarrow$ Theorem 5

$$* |A^T| = |A|$$

1-If A has a row or column consisting entirely of zeros, then $|A|=0$.

2-If A has two identical rows or two identical columns, then $|A|=0$.

1) $\begin{bmatrix} 1 & 2 & 4 & 6 \\ 5 & -1 & 0 & 4 \\ 6 & 7 & -9 & 10 \\ 1 & 2 & 4 & 6 \end{bmatrix} \Rightarrow$ $\therefore |A|=0$

2) $\begin{bmatrix} 1 & 2 & 3 & 4 \\ 6 & 9 & 10 & 20 \\ 0 & 0 & 0 & 0 \end{bmatrix} \Rightarrow$ $|A|=0$

2.2: Prosperities of determinants:

Row operations :

1- $|B| = -|A| \Rightarrow A = \begin{bmatrix} 2 & 4 \\ 6 & 5 \end{bmatrix} \Rightarrow B = \begin{bmatrix} 6 & 5 \\ 2 & 4 \end{bmatrix} \Rightarrow$ بس تبديل صفين أو عمودين بالتبديل.

2- $|B| = K|A| \Rightarrow A = \begin{bmatrix} 1 & 2 \\ 4 & 5 \end{bmatrix} \Rightarrow B = \begin{bmatrix} 2 & 4 \\ 4 & 5 \end{bmatrix} = B = 2A$.

3- $|B| = |A| \Rightarrow A = \begin{bmatrix} 1 & 4 \\ 5 & 5 \end{bmatrix} \Rightarrow B = \begin{bmatrix} 1 & 4 \\ 0 & -25 \end{bmatrix} \Rightarrow |B| = -25 = |A|$.
 $-5R_1 + R_2$

Let E be an elementary matrix then :

$|E| = 1 - (-1)$ if E is of type 1.

= $2 - k \neq 0$ if E is of type 2. $\Rightarrow |E|$ هو عدد القسمة

= $3 - (1)$ if E is type of 3.

* $|kA| = k^n |A|$.

* $|I| = 1$

* $|EA| = |E| |A|$.

* If A and B matrices then $\Rightarrow |AB| = |A| |B|$.

* If A is singular, then $|A| = 0$ and so AB is singular, therefore $|AB| = 0 = |A| |B|$.

$|A+B| \neq |A| + |B|$

* لا يمكن تبسيطها في الجمع

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

* If $A^T A = I$, then $|A| = \pm 1 \Rightarrow |A^T A| = |I|$

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

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

* If he told me to use elimination Method to evaluate $\det(A)$.

REF

2.3 : Additional topics and applications :

The adjoints of a matrix:

$$\text{adj}(A) = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1n} \\ A_{21} & A_{22} & \dots & A_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ A_{n1} & A_{n2} & \dots & A_{nn} \end{bmatrix}^T$$

$$* A_{ij} = (-1)^{i+j} |M_{ji}|.$$

$$* A \text{adj}(A) = |A| I_n.$$

$$* |\text{adj}(A)| = |A|^{n-1}.$$

$$* A^{-1} = \frac{1}{|A|} \text{adj}(A).$$

$$* |\text{adj}(A)| = |A|^{-1} |A| = \text{adj}(A^{-1}).$$

$$* \text{adj}(\text{adj}(A)) = |A|^{n-2} A.$$

$$* \text{Cramer's Rule} \Rightarrow X_i = \frac{|A_i|}{|A|}, \quad i=1, 2, \dots, n.$$

example $\Rightarrow$ $\begin{cases} x_1 + 2x_2 + x_3 = 5 \\ 2x_1 + 2x_2 + x_3 = 6 \\ x_1 + 2x_2 + 3x_3 = 9 \end{cases} \quad A = \begin{vmatrix} 1 & 2 & 1 \\ 2 & 2 & 1 \\ 1 & 2 & 3 \end{vmatrix}$

$$|A| = \begin{vmatrix} 5 & 2 & 1 \\ 6 & 2 & 1 \\ 9 & 2 & 3 \end{vmatrix} \Rightarrow \text{عوضنا الناتج في المعادلات}$$

$$\rightarrow x_1 = \frac{|A_1|}{|A|} \quad \text{و هكذا}$$

* Evaluate the following determinant, write your answer as a polynomial in x :-

$$\begin{vmatrix} a-x & b & c \\ 1 & -x & 0 \\ 0 & 1 & -x \end{vmatrix}$$

$$\rightarrow |A| = (a-x) \begin{vmatrix} -x & 0 \\ 1 & -x \end{vmatrix} - b \begin{vmatrix} 1 & 0 \\ 0 & -x \end{vmatrix} + c \begin{vmatrix} 1 & -x \\ 0 & 1 \end{vmatrix}$$

$$= (a-x)x^2 - b(-x) + c$$

$$= -x^3 + ax^2 + bx + c.$$

* If A is singular then $A \text{adj}(A) = ??$

$A \text{adj}(A) = |A| I$ if A is singular $|A| = 0$ so $\text{adj}(A) = 0$.

3.1: Definitions and examples:

V is a vector space is a set of elements together with the operations of addition and scalar multiplication :-

C1: if $x \in V$ and α is scalar, then $\alpha x \in V$ "closed under scalar multiplication"

C2: if $x, y \in V$, then $x + y \in V$ "closed under addition".

A1: $x + y = y + x$.

A2: $(x + y) + z = x + (y + z)$.

A3: $\exists$ an element $0 \in V$ such that $x + 0 = 0 + x = x, \forall x \in V$.

A4: $\forall x \in V, \exists -x \in V$ such that $x + -x = 0$.

A5: $\alpha(x + y) = \alpha x + \alpha y$.

A6: $(\alpha + \beta)x = \alpha x + \beta x$.

A7: $(\alpha\beta)x = \alpha(\beta x)$.

A8: $1x = x$.

$\Rightarrow$ Notation $(V, +, \cdot)$

$\Rightarrow V = \mathbb{R}^2 \Rightarrow \mathbb{R}^2$ with usual addition and multiplication is a vector space.

2 $V = \mathbb{R}^2$ with usual $+$ and $\cdot$ is a vector space where $(a, b) + (c, d) = (a + c, b + d)$
 $\alpha(a, b) = (\alpha a, \alpha b)$.

3 $M_{m \times n} = \mathbb{R}^{m \times n}$ is the set of all $m \times n$ matrices with real entries under addition and scalar multiplication is a vector space.

4 The set of all real valued functions under $+$ and $\cdot$:-
 $\begin{cases} (f + g)(x) = f(x) + g(x). \\ (\alpha f)(x) = \alpha f(x). \end{cases}$

5 $C[a, b] := \{f: [a, b] \rightarrow \mathbb{R} : f \text{ is continuous on } [a, b]\}$
 $\begin{cases} (f + g)(x) = f(x) + g(x). \\ (\alpha f)(x) = \alpha f(x). \end{cases}$

$n \rightarrow \infty$

$$\boxed{1} \quad C[a, b] = \{ f: [a, b] \rightarrow \mathbb{R} : f^{(n)} \text{ is continuous on } [a, b] \}$$

$$(f+g)(x) = f(x) + g(x).$$

$$(\alpha f)(x) = \alpha f(x).$$

$$\boxed{2} \quad P_n = \{ f(x) = a_{n-1} x^{n-1} + \dots + a_1 x + a_0 \}$$

$$(f+g)(x) = f(x) + g(x)$$

$$(\alpha f)(x) = \alpha f(x).$$

* irrationales $= (\sqrt{2}, \sqrt{3}, e, \pi, \dots)$ is not a vector space.

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

* $U = \{ f(x) = \deg(f) = 3 \}$ is not a vector space.

$$\hookrightarrow 1-x^3, 1+x^3 \in U \quad \text{but} \quad 1-x^3 + 1+x^3 = 2 \notin \deg=3.$$

* Let V be a vector space then:-

$$\boxed{1} \quad 0 \cdot v = \vec{0}, \forall v \in V.$$

$$\boxed{2} \quad \text{if } x+y = \vec{0}, \text{ then } y = -x.$$

$$\boxed{3} \quad -1 \cdot v = -v, \forall v \in V.$$

3.2: Subspace and spanning sets:

* Subspace $\Rightarrow$ ① $x+y \in S$.

② $\alpha x \in S$.

③ $0 \in S$, $S \neq \emptyset$.

* Let S and T be subspace of vector space V then:-

① $S \cap T$ is subspace

② $S \cup T$ is not always a subspace of V .

③ $S+T$ is subspace of V .

* Nullspace $\Rightarrow Ax=0$ (REF) $\Rightarrow$ is a subspace of $\mathbb{R}^n$.

طرق free variation $\leftarrow$ ١- المتغيرات الحرة $\Rightarrow$ طريقة إجابة.
٢- بإستخدام المصفوفة.

* Linear combination:- $c_1 u_1 + c_2 u_2 + \dots + c_n u_n \Rightarrow$ is a subspace of V .

$\hookrightarrow \text{span}(u_1, u_2, \dots, u_k)$.

example $\Rightarrow u = \begin{pmatrix} 2 \\ 3 \end{pmatrix}$, $u_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$, $u_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \Rightarrow V = c_1 u_1 + c_2 u_2 \Rightarrow \begin{cases} c_1 + c_2 = 2 \\ c_2 = 3 \end{cases} \Rightarrow \begin{cases} c_1 = -1 \\ c_2 = 3 \end{cases}$.

$\therefore V = -u_1 + 3u_2$ is a linear combination of $\text{span}(u_1, u_2)$.

* spanning set:- if $V = \text{span}(u_1, \dots, u_k)$ $\Rightarrow$ consistent $\Rightarrow$ $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$.

$\hookrightarrow e_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$, $e_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$, $e_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$.

* $\{e_1, e_2, \dots, e_n\}$ is the standard spanning set for $\mathbb{R}^n$.

$$x = x_1 \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} + \dots + x_n \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}$$

* $\{1, x, x^2, x^3\}$ is a spanning set for P_4 .

* standard spanning for $\mathbb{R}^{2 \times 2}$ are:-

$$\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}.$$

* Let A be a 4×3 matrix and let $b \in \mathbb{R}^4$. How many possible solutions could the system $Ax=b$ have if $N(A)=\{0\}$? Answer the same question in the case $N(A) \neq \{0\}$. Explain your answer.

↳ [a] If $N(A)=0$ then $y=x_0 + \vec{0} \rightarrow 0$ so $y=x_0$ Therefore, $Ax=b$ has no solution or a unique solution.

↳ [b] If $N(A) \neq 0$ then $y=x_0 + \vec{z} \rightarrow \neq 0$ so $y \neq x_0$ Therefore $Ax=b$ has no solution or infinite solutions.

* Let $\{x_1, x_2, \dots, x_k\}$ be a spanning set for a vector space V .

↳ if we add another vector x_{k+1} to the set, will we still have a spanning set? $\Rightarrow$ Yes.

* if we delete one of the vector say x_k from the set, will we still have a spanning set $\Rightarrow$ No.

$\{(1,0,0), (0,1,0), (0,0,1)\} \Rightarrow$ spans to $\mathbb{R}^3$ but

$\{(1,0,0), (0,1,0)\}$ does not.

3.3: linear independence:

⇒ We called about V is linearly independent:-

$$\hookrightarrow c_1 v_1 + c_2 v_2 + \dots + c_n v_n = 0 \Rightarrow \boxed{c_1 = c_2 = \dots = c_n = 0}$$

other wise, they are linearly dependent.

⇒ Is $\begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ linearly independent??

$$c_1 \begin{pmatrix} 1 \\ 1 \end{pmatrix} + c_2 \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow 0 = c_2, c_1 = 0 \text{ (بطل)}$$

⇒ R.E.F.

*If the system was underdetermined homogeneous $\Rightarrow$ it has infinite solution.
⇒ so it is always linearly dependent.

Theorem 1:- A set of vectors in $\mathbb{R}^n$ are linearly independent iff the matrix $A = [v_1 \ v_2 \ \dots \ v_n]$ is nonsingular.

⇒ Remark $\Rightarrow |A| \neq 0 \Rightarrow$ linearly independent.

For $n \times n$ matrix is Theorem 1 is b

$\Leftrightarrow P_n$ of Eigen values

$$\text{Ex} \Rightarrow P_1(x) = 2x^2 + x + 8, \quad P_2(x) = x^2 + 8x + 7, \quad P_3(x) = x^2 - 2x + 3.$$

$$\hookrightarrow c_1 P_1(x) + c_2 P_2(x) + c_3 P_3(x) = 0$$

$$c_1 (2x^2 + x + 8) + c_2 (x^2 + 8x + 7) + c_3 (x^2 - 2x + 3) = 0.$$

⇒ x^2 terms, x terms, constant terms $\Rightarrow$ parallel

$$2c_1 + c_2 + c_3 = 0$$

$$c_1 + 8c_2 - 2c_3 = 0$$

$$8c_1 + 7c_2 + 3c_3 = 0$$

$\Rightarrow$ values, P_1, P_2, P_3
Theorem 1

Theorem 2:- A set of vectors are linearly dependent iff one of them is a linear combination of the remaining set of vectors.

⇒ example = $\left\{ \begin{pmatrix} 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}$ are linearly dependent
 $v_1 \quad v_2 \quad v_3$

since $v_3 = v_1 + v_2 \Rightarrow (v_3 \text{ is a linear combination of } v_1 \text{ and } v_2).$

Theorem 13: - A set of vectors are linearly independent iff every vector is uniquely written as a linear combination of $v_1, v_2, \dots, v_n$.

* the vector space $C^n[a, b]$.

$W(f_1, f_2, \dots, f_n)$ on $[a, b]$ by:

$$W(f_1, \dots, f_n) = \begin{vmatrix} f_1 & \dots & f_n \\ f_1' & \dots & f_n' \\ \vdots & \dots & \vdots \\ f_1^{(n-1)} & \dots & f_n^{(n-1)} \end{vmatrix}$$

← Wronskian of $f_1, \dots, f_n$

if $W(f_1, f_2, \dots, f_n)(x_0) \neq 0$ then $\{f_1, \dots, f_n\}$ are linearly independent.

* If $f_1, f_2, \dots$ are linear dependent then $W(f_1, \dots, f_n) = 0$.

if $W(f_1, f_2, \dots, f_n) = 0$ we can't say about anything $\Rightarrow$ $c_1 v_1 + c_2 v_2 + \dots + c_n v_n = 0$

* If we add a vector x_{k+1} to set $\{x_1, x_2, \dots, x_k\}$, the set $\{x_1, x_2, \dots, x_k, x_{k+1}\}$ may or may not be a linear independent set.

* If we delete x_k from $\{x_1, x_2, \dots, x_k\}$, it will still be linear independent.

* any nonempty subset of a linearly independent set of vectors $\{v_1, \dots, v_n\}$ is also linearly independent.

* Let A be an $m \times n$, show that if A has linearly independent column vector, then $N(A) = \{0\}$.

$\Rightarrow$ linearly dependent $\Rightarrow c_1 x + c_2 y = 0$, where c_1, c_2 are not $\neq 0$.

$$\hookrightarrow x = -\frac{c_2}{c_1} y$$

$\Rightarrow$ both vectors (x and y) placed at the origin, they will lie along the same line, but if they are linearly independent will not lie at the same line.

If the matrix was not square we can not use the det:

1- we use homogeneous

2-REF if x involve free variables, if they are not trivial solution, then the vectors are linearly dependent , if there are no free variables then $c=0$ is the only solution and hence the vectors must be linearly independent.

3.4: basis and dimension:

1) $\downarrow$ span V

2) Linearly independent

* The zero vector space $\{0\}$ has dimension zero basis $\emptyset$.

1) $\dim \mathbb{R}^n = n$. 2) $\dim \mathbb{P}_n = n$. 3) $\dim \mathbb{R}^{m \times n} = m \cdot n$. 4) $\dim \{0\} = 0$.

5) $\dim \mathbb{R} = 1$. 6) $\dim C^{\infty}[a, b] = \infty$.

Theorem 1:- Let $\{v_1, v_2, \dots, v_n\}$ be a spanning set for V . If $w_1, w_2, \dots, w_k \in V$, $k > n$, then $w_1, w_2, \dots, w_k$ are linearly dependent.

$\rightarrow$ example $\rightarrow \left\{ \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 4 \\ 3 \end{pmatrix} \right\}$ is a spanning set for $\mathbb{R}^2$
 $\left\{ \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 4 \\ 3 \end{pmatrix}, \begin{pmatrix} -7 \\ 3 \end{pmatrix} \right\}$ are linearly dependent.

Theorem 2:- Let $\{v_1, \dots, v_n\}$ and $\{w_1, \dots, w_k\}$ be two bases for a vector space V . Then

Theorem 3:- Let V be a vector space with $\dim V = n$ then the following are equivalent:

i) $\{v_1, \dots, v_n\}$ is a basis.

ii) $\{v_1, \dots, v_n\}$ span V .

iii) $\{v_1, \dots, v_n\}$ linearly independent.

Summary $\Rightarrow \dim V = n$ then:-

i) A set $v_1, \dots, v_k$, $k > n$ linearly dependent.

ii) A set $v_1, \dots, v_k$, $k < n$ can not span V .

iii) If $k = n$ and $v_1, \dots, v_k$ are linearly independent or span V , then $\{v_1, v_2, \dots, v_k\}$ is a basis for V .

iv) A spanning set of $v_1, v_2, \dots, v_k$, $k > n$ can be reduced (piled down) to a basis for V .

v) A linearly independent set $v_1, \dots, v_k$, $k < n$ can be extended to a basis for V .

$\rightarrow$ Examples:-

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 0 \end{bmatrix} \right\}$$

Linearly independent
but not span for $\mathbb{R}^3$

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix} \right\}$$

$\downarrow$
A basis for
 $\mathbb{R}^3$

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}, \begin{bmatrix} 7 \\ 5 \\ 9 \end{bmatrix} \right\}$$

spans $\mathbb{R}^3$, but
it is linearly dependent.

* find a basis and dimension ϕ :-

$$S = \{ (a+3b+c, 2a+6b+c)^T : a, b, c \in \mathbb{R} \}.$$

$$a \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} + b \begin{pmatrix} 3 \\ 6 \\ 0 \end{pmatrix} + c \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}.$$

Notice that $v_2 = 3v_1 \Rightarrow$ the $\{v_1, v_2, v_3\}$ linearly dependent.

$\therefore S = \text{span} \left\{ \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \right\} \rightarrow$ basis for S .

linearly independent:

dimension = 2.

$\Rightarrow$ The standard basis for $\mathbb{R}^3$ is $\{e_1, e_2, e_3\}$.

$\Rightarrow$ The standard basis for $P_n = \{1, x, x^2, \dots, x^{n-1}\}$.

$\Rightarrow$ since they are linearly independent, they form a basis for:-

$N(A) = \square$ RREF.

[2] بما لا يقل عن

[3] بما لا يقل عن free variables

$N(A) \rightarrow$

$\Rightarrow$ If V is a vector space of dimension $n > 0$, then:-

[1] any set of n linearly independent vector space V .

[2] any n vectors that spans V are linearly independent.

3.5: change of basis:

* $E = \{u_1, u_2, \dots, u_n\}$ a basis of V .

then any $v \in V$ can be written uniquely as:-

$$V = \alpha u_1 + \alpha u_2 + \alpha u_3 + \dots + \alpha u_n$$

$\vec{\alpha} = (\alpha_1, \alpha_2, \dots, \alpha_n)$ coordinate of v with respect to a basis E .

$$\downarrow [v]_E \text{ or } v_E \Rightarrow [v]_E = \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{bmatrix}$$

* If E is a standard basis for a vector space R then $[v]_E = v, \forall v \in R^n$.

* U is called transition matrix from the basis:-

$$E = \{u_1, u_2\}$$

$$[u_1, u_2] \xrightarrow{U} [e_1, e_2]$$

$$S = [U_2^{-1} U_1]$$

$$\downarrow [U_2^{-1}]$$

$$\rightarrow [u_1, u_2]$$

$$* v_B = P_{B' \rightarrow B} [v]_{B'}, v_{B'} = P_{B \rightarrow B'} [v]_B$$

$\Rightarrow$ Transition matrices $\Rightarrow P_{B' \rightarrow B}$

from B' to B .

$$P_{B \rightarrow B'} \Rightarrow [B' | B] \text{ RREF}$$

$$\downarrow [I | P_{B \rightarrow B'}]$$

$$, P_{B' \rightarrow B} \Rightarrow [B | B'] \text{ RREF}$$

$$\downarrow [I | P_{B' \rightarrow B}]$$

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} \Rightarrow [2x-1, 2x+1]$$

= transition: $\leftarrow \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ $\leftarrow \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ $\leftarrow \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

$$x = c_1 (2x-1) + c_2 (2x+1) \Rightarrow c_1, c_2 \text{ عاين}$$

$$1 = c_1 (2x-1) + c_2 (2x+1) \Rightarrow c_1, c_2 \text{ عاين}$$

* Let V be a finite dimensional vector space with $\dim V = n$, if $E = [v_1, v_2, \dots, v_n]$, $F = [w_1, w_2, \dots, w_n]$ be two basis.

Then the transition matrix from the basis E into the basis F is the $n \times n$ nonsingular matrix.

$$\square [v_1 + v_2 + \dots + v_k]_E = [v_1]_E + [v_2]_E + \dots + [v_k]_E.$$

$\square v_1, v_2, \dots, v_k$ are linearly independent iff $[v_1]_E, \dots, [v_k]_E$ are linearly independent.

$$* [x]_F = S_{F \rightarrow E} [x]_E = (U_2^{-1} U_1) [x]_E.$$

$$\hookrightarrow x = \alpha_1 x_1 + \alpha_2 x_2.$$

$$[x]_E = \begin{bmatrix} \alpha_1 \\ \alpha_2 \end{bmatrix}$$

3.6: row space and column space :

basis for the nullspace $\Rightarrow$ 1- REF. , 2- إبطالها بالفر. , 3- $x_1, x_2, \dots$ بطالع قيم

□ Row space $\Rightarrow$ 1. REF $\Rightarrow$ $\perp$ line

$\rightarrow R(U) = R(A) = \text{the nonzero rows}$ is a basis for $R(A)$ قِيمَ لِيَاوِي U هِيَ قِيمَ لِيَاوِي A
 Row space $\rightarrow$ قِيمَ لِيَاوِي A

بالتعميم: $L \neq \text{القيم في } A$ في نفس الترتيب $\Rightarrow$ مرتبة $A \Rightarrow$ I-REF $\Rightarrow$ column space $\Rightarrow$ is a column space. A يكون الأول عودون في column space.

③ The nullity $\Rightarrow \text{Null}(A) = \dim N(A)$

□ Rank of $A \Rightarrow \text{rank}(A) = \dim(A)$.

* $\text{Rank}(A) + \text{Null}(A) = n$, $\Rightarrow \text{rank} + \dim(N(A)) = n$.

إني أكتب القويح أو أي عدد ما دخل الـ Row space كامل معادلة وهذا يسمى موارد $\Rightarrow$ dependency relation. ويسمى الـ A.

* If $m \times n$ matrix, the dimension of the row space of A = the dimension of the column space of A . $\rightarrow A_{3 \times 2}$, rank = 2.

① $Ax=b$ is consistent iff b is in the column space of A .

② $Ax=b$ is consistent iff the column vectors of A span $\mathbb{R}^m$.

→ The $Ax=b$ has at most one solution iff A is linearly independent

③ $n \times n$, A is nonsingular iff the column vector of A form a basis for $\mathbb{R}^n$.

$$\rightarrow x = A^{-1}b$$

* إذا طلب ال dimension يعني بطلع عدد row space أو column space

. REF 

dimension: 1×2 row space. $\rightarrow$ $\{ \vec{a}, \vec{b} \}$ basis

* إذا يكون معطين A و b و b أحسن إذا هي consistent و b هي حل لا لها $b \in [A]$

$$x_1 a_1 + x_2 a_2 = b$$

دکلم إذا طلع عندي أجوبة يكون $x_1 \begin{pmatrix} 1 \\ 2 \end{pmatrix} + x_2 \begin{pmatrix} 2 \\ 4 \end{pmatrix} = \begin{pmatrix} 4 \\ 6 \end{pmatrix} \Rightarrow$ ما طلع يكون

* How many solutions will the linear system $Ax=b$ have if b is in the column space of A and the column vectors of A are linearly dependent.

↳ If $b \in C(A)$, then the system is consistent. If the column vectors of A are linearly dependent, then the system will have infinitely many solutions.

* Let A be an $m \times n$ matrix with $m < n$, Let $b \in \mathbb{R}^m$ and suppose that $N(A) = \{0\}$.

□ What can you conclude about the column vectors of A ? Are they linearly independent? Do they span $\mathbb{R}^m$? Explain

↳ $N(A) = \{0\} \Rightarrow \text{Nullity} \Rightarrow \text{rank}(A) = n$. A has n columns and the dimension of its column space $= n \Rightarrow$ the columns are linearly independent. Dimension of $\mathbb{R}^m = m$ and $m < n$. Therefore, columns of A can't span $\mathbb{R}^m$.

□ How many solutions will the system $Ax=b$ have if b is not in the column space of A ? How many solutions will there be if b is in the column space of A ?

↳ If b doesn't belong in the column space of A , system $Ax=b$ is inconsistent, and has no solutions. If b belongs in the column space, $Ax=b$ is consistent and has infinitely many solutions.

* Let A be 4×4 matrix with RREF given by $\Rightarrow$

$$U = \begin{bmatrix} 1 & 0 & 2 & 1 \\ 0 & 1 & 1 & 4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \text{ if } a_1 = \begin{bmatrix} -3 \\ 5 \\ 2 \\ 1 \end{bmatrix} \text{ and } a_2 = \begin{bmatrix} 4 \\ -3 \\ 7 \\ -1 \end{bmatrix} \text{ find } a_3 \text{ and } a_4.$$

$$\begin{aligned} u_3 &= 2u_1 + u_2 \\ a_3 &= 2a_1 + a_2 \\ \text{V.P.S.} & \end{aligned} \quad \left\{ \begin{aligned} u_4 &= u_1 + 4u_2 \\ a_4 &= a_1 + 4a_2 \\ \text{V.P.S.} & \end{aligned} \right.$$

* A and B are $n \times n$ matrices $N(A-B) = \mathbb{R}^n$ then $A=B$.

* $AB=0$ iff the column space of B is a subspace of null space of A .

* $AB=0$ then the sum of the ranks of A and B cannot exceed n .

4.1 Linear Transformation iff $L: V \rightarrow W$
 1. $L(u_1 + u_2) = L(u_1) + L(u_2)$, $\forall u_1, u_2 \in V$.
 2. $L(\alpha u) = \alpha L(u)$, $\forall u \in V, \forall \alpha \in \mathbb{R}$.

$\Rightarrow$ If $V = W$ then $L: V \rightarrow V$ is said to be linear operator.

$\Rightarrow L: V \rightarrow W$ is a linear Transformation iff :-
 $L(\alpha u_1 + \beta u_2) = \alpha L(u_1) + \beta L(u_2)$.

$\Rightarrow$ Kernel and Images (Range):

$$L(u) = 0_w$$

إذا طبع الجواب صفر

one-one

dim kernel:-

يعود الصورة إلى معطياتها
 بالسؤال بأولي بالصفر وبولي
 القيم بعد ما بريمع للأصل نعوضها
 بـ $L(u)$ وللي قبل جزئه عامل
 متفرقة

$$L: W = L(u)$$

2) image وطبع على

بكونه الجواب onto.

$$L \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 \\ 0 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

هنا هو الـ image

Example:- $L \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

لـ بتوف صورة بكونه

أها بالسؤال

$$x_1 = 0 \Rightarrow \begin{pmatrix} x_1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$L \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ x_2 \end{pmatrix} = x_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

هنا kernel

[6.1] $\Rightarrow$ Eigen values and Eigen vectors

$\downarrow$

λ

$\downarrow$

$$A\vec{v} = \lambda \vec{v}$$

$$\Rightarrow \lambda \text{ قيمة } \Rightarrow |A - \lambda I_n| = 0$$

$$\Rightarrow N(A - \lambda I) \neq \{0\}, \quad A - \lambda I \text{ is singular.}$$

$$\begin{vmatrix} 3-\lambda & 2 \\ 3 & -2-\lambda \end{vmatrix} \Leftarrow \text{main diagonal } \lambda \text{ في الـ } A$$

دوم المحدد ونعلم القيمة ودوم قيمة الـ λ

$\Leftarrow$ هسة الـ Eigen values $\Rightarrow$ إذا طلوعا فتلحق القيم يعني $\lambda_1 = 4, \lambda_2 = -3$

$\Leftarrow$ هو من قيمة الـ λ في المدة التي طرقت مالاً :

$$\begin{vmatrix} 3-4 & 2 \\ 3 & -2-4 \end{vmatrix} = \begin{vmatrix} -1 & 2 \\ 3 & -6 \end{vmatrix}$$

$$\begin{vmatrix} -1 & 2 & 0 \\ 3 & -6 & 0 \end{vmatrix} \Leftarrow N(A - \lambda I)$$

(ARREF) $\Leftarrow$ دوم الـ Null الطبيعي يكون هاد الـ $\vec{v}_1$ ونجوم الـ $\lambda_2 = -3$

$\Rightarrow$ هسة إذا طلوعا الـ فتلحق ونفس القيمة دوم الـ Null

$$\Rightarrow \text{complex eigen value} \Rightarrow z = \alpha + \beta i \Rightarrow \bar{z} = \alpha - \beta i$$

$\Leftarrow$ conjugate (المرافق)

$$\lambda = \pm 2i \Leftarrow \lambda^2 + 4 = 0 \quad \text{كل هاد مالاً إذا اطاق}$$

$$\lambda_2 = 2 - i, \quad \lambda_1 = 2 + i \quad \Leftarrow \text{لفترخ عندي}$$

$\Leftarrow$ ابدى اول الـ $\vec{v}$ دوم لوصة والثانية

$$\vec{v}_2 = \begin{pmatrix} -1 \\ 5 \end{pmatrix}$$

$$\vec{v}_1 = \begin{pmatrix} 1 \\ 5 \end{pmatrix} \Leftarrow \text{لطاق}$$

$$A\vec{v} = \lambda \vec{v}$$

$$-i^2 = 1 \Rightarrow i^2 = -1$$

Trace denoted $\text{tr}(A)$: is the sum of all entries on the main diagonal.

$$\Rightarrow |A| = \lambda_1 \lambda_2 \dots \lambda_n \quad ? \quad \text{يعني } \lambda \text{ هو } 0 \text{ يعني } A \text{ singular}$$
$$\Rightarrow \text{tr}(A) = \lambda_1 + \lambda_2 + \dots + \lambda_n \quad |A - \lambda I| = 0$$

Let A be an $n \times n$ matrix, then A is singular if $\lambda = 0$ is an eigenvalue of A .

Let A be an $n \times n$ matrix, then A and A^T have the same eigenvalue.

$$\Rightarrow A^n v = \lambda^n v \Rightarrow Av = \lambda v$$

Eigenvalue for triangular : will be the main diagonal. $= \lambda$.

Let A be an $n \times n$ nonsingular matrix. If λ is an eigenvalue of A , then $\frac{1}{\lambda}$ is an eigenvalue for A^{-1} with the same eigenvectors. $\Rightarrow A^{-1}v = \frac{1}{\lambda}v$.

