

LINEAR ALGEBRA

day / date:

the branch of mathematics concerned with linear equations, matrices, determinants, vector spaces, etc.

LINEAR → RELATING TO THE FIRST DEGREE; HAVING NO VARIABLE RAISED TO ANY POWER

LINEAR EQUATIONS: A line in the xy -plane is an equation of the form

$$a_1x + a_2y = b \quad (1)$$

Equation (1) is called linear equation in the variables x and y . We define a linear equation in the n variables x_1, x_2, \dots, x_n to be one that can be expressed in the form

$$a_1x_1 + a_2x_2 + \dots + a_nx_n = b \quad (2)$$

where a_1, a_2, \dots, a_n and b are REAL CONSTANTS.

The variables in a linear equation are called the unknowns and in (2) unknowns are

$$x_1, x_2, x_3, \dots, x_n$$

the real constants a_1, a_2, \dots, a_n are also called coefficients

* variables should have no power

* variables do not appear as arguments for trigonometric, logarithmic, or, exponential functions.

* do not indulge any products or roots of variables



day / date:

A general system of m linear equations in n unknowns will be written

$$\left. \begin{array}{l} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \vdots \quad \vdots \quad \vdots \quad \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m \end{array} \right\} \quad (3)$$

the first subscript i ($1 \leq i \leq m$) on the coefficient a_{ij} indicates the equation in which the coefficients occurs and the second subscript j ($1 \leq j \leq n$) indicates which unknown it multiplies
linear system (3) can be abbreviated by writing only the rectangular array of numbers.

$$\left[\begin{array}{cccc|c} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & b_2 \\ \vdots & \vdots & & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} & b_m \end{array} \right]$$

This is called **AUGMENTED MATRIX** due to **b_i 's** ($1 \leq i \leq m$)
AUGMENT → increase/enlarge due to presence of the entries on RHS of the linear system (3).

* if a system of equations that has **NO SOLUTION** is said to be **inconsistent**. If there is **at least one solution** then it is **CONSISTENT**.



MATRICES

day / date:

MATRIX: A matrix is a rectangular array of numbers enclosed in brackets. The numbers in brackets may be called entries in the matrix. (2)

The size of a matrix is described by specifying the number of rows (horizontal lines) and columns (vertical lines).

Size : row x column

↪ if a matrix has $\text{row} = \text{column}$ then it is a **SQUARE MATRIX**

↪ if a matrix has only one column is also called a **COLUMN VECTOR**. and similarly a matrix which has only one row is called a **ROW VECTOR**.

→ the entries $a_{11}, a_{22}, \dots, a_{nn}$ in a square matrix A of order n , are said to be on the main diagonal of A as shown here in following matrix

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$

DIAGONAL OF
 A .

→ a square matrix in which all entries off the main diagonal are zero is called **DIAGONAL MATRIX**

$$\begin{bmatrix} 2 & 0 \\ 0 & -1 \end{bmatrix}$$

→ if all entries are zero it is called a zero or null matrix, denoted by 0.

✓ if A and B are any two matrices of same size then the sum A+B is the matrix obtained by ADDING together the corresponding entries in two matrices

$$A = \begin{bmatrix} 2 & 1 & 0 \\ -1 & 0 & 2 \\ 4 & -2 & 7 \end{bmatrix}, B = \begin{bmatrix} 3 & 2 \\ 5 & 6 \end{bmatrix}, C = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Find A+B and B+C.

SOLUTION A+B is NOT defined since A & B are of different size.

$$B+C = \begin{bmatrix} 3+1 & 2+0 \\ 5+0 & 6+1 \end{bmatrix} \Rightarrow \begin{bmatrix} 4 & 2 \\ 5 & 7 \end{bmatrix}$$

By the definition of EQUAL MATRICES the linear system

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

$$\vdots \quad \vdots \quad \vdots \quad \vdots$$

$$a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m$$

can be written as

$$\left[\begin{array}{l} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n \\ \vdots \quad \vdots \quad \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n \end{array} \right] = \left[\begin{array}{l} b_1 \\ b_2 \\ \vdots \\ b_m \end{array} \right] \quad \text{↑ } m \times 1$$

further the $m \times 1$ matrix on the left hand side can be written as the product of two matrices given by

$$\left[\begin{array}{cccc} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{array} \right] \left[\begin{array}{l} x_1 \\ x_2 \\ \vdots \\ x_n \end{array} \right] = \left[\begin{array}{l} b_1 \\ b_2 \\ \vdots \\ b_m \end{array} \right] \quad \text{↑ } n \times 1 \quad \text{↑ } m \times 1$$

$\curvearrowleft m \times n \text{ matrix}$

if we designate these matrices by A , X and B , respectively, the original system of m equations in n unknowns has been replaced by the single matrix equation.

$AX = B$, where A is called the coefficient matrix. But note the following

$$A \cdot X = B$$

$\overset{m \times n}{\underset{\text{OUTSIDE}}{\boxed{\text{INSIDE}}}} \quad \overset{n \times 1}{\underset{\text{OUTSIDE}}{\boxed{\text{INSIDE}}}} \quad \overset{m \times 1}{\underset{\text{OUTSIDE}}{\boxed{\text{INSIDE}}}}$

No of columns of $A = n$
No of rows of $X = n$

✓ Subject
 → Two matrices A and B can be multiplied if the number of columns of A and are equal to the number of rows of B .

$$\begin{matrix} A & B \\ \text{m} \times r & r \times n \end{matrix} = AB \quad \begin{matrix} m \times n \end{matrix}$$

INSIDE

OUTSIDE

Find AB , where 2×3 3×4

$$A = \begin{bmatrix} 1 & 2 & 4 \\ 2 & 6 & 0 \end{bmatrix}, B = \begin{bmatrix} 4 & 1 & 4 & 3 \\ 0 & -1 & 3 & 1 \\ 2 & 7 & 5 & 2 \end{bmatrix}$$

Now the product AB is 2×4 matrix

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

$$= \begin{bmatrix} C_{11} & C_{12} & C_{13} & C_{14} \\ C_{21} & C_{22} & C_{23} & C_{24} \end{bmatrix} = C \text{ (SAY)}$$

where

$$C_{11} = 1(4) + 2(0) + 4(2) = 12$$

$$C_{12} = 1(1) + 2(-1) + 4(7) = 27$$

$$C_{13} = 1(4) + 2(3) + 4(5) = 30$$

then

$$AB = \begin{bmatrix} 12 & 27 & 30 & 13 \\ 8 & -9 & 26 & 12 \end{bmatrix}$$

Two matrices A and B commute if $AB = BA$ but in general $AB \neq BA$

→ if A and B are two square matrices of same size, then find the condition such that

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

NOTE: in such type of questions don't use matrix entries. Just use matrix symbols.

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

$$= A^2 + \cancel{AB} + \cancel{BA} + B^2$$

↪ if A and B commute then

$$AB = BA$$

$$\therefore (A+B)^2 = A^2 + 2AB + B^2$$



→ if A, B and C are matrices such that AB and BC are defined then $A(BC) = (AB)C$, which is also called **ASSOCIATE LAW** for multiplication.

→ If A is $m \times n$ matrix and D is diagonal matrix of order m then find the multiplication rule for DA .

Solution: DA

$$D = \begin{bmatrix} d_1 & 0 & \cdots & 0 \\ 0 & d_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & \cdots & d_m \end{bmatrix} \quad A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}$$

$$= \begin{bmatrix} d_1 a_{11} & d_1 a_{12} & \cdots & d_1 a_{1n} \\ d_2 a_{21} & d_2 a_{22} & \cdots & d_2 a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ d_m a_{m1} & d_m a_{m2} & \cdots & d_m a_{mn} \end{bmatrix}$$

PROOF

(1) Show that if AB and BA are both defined, then AB and BA are square matrices.

$$A_{m_1 \times n_1} \quad \& \quad B_{m_2 \times n_2}$$

$$n = n_1 = m_2 \rightarrow (A_{m_1 \times n_1} \text{ is } B_{n \times n_2})$$

$$m_1 = n_2 = m$$

$$A_{m \times n} \quad B_{n \times m} = AB_{m \times m} \quad \text{this is square matrix}$$

LINEAR COMBINATIONS

We showed in Example 5 that

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

$$AB = \begin{bmatrix} 12 & 27 & 30 & 13 \\ 8 & -4 & 26 & 12 \end{bmatrix}$$

The column matrices of AB can be expressed as linear combinations of column matrices of A as follows.

$$\begin{bmatrix} 12 \\ 8 \end{bmatrix} = 4 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 0 \begin{bmatrix} 2 \\ 6 \end{bmatrix} + 2 \begin{bmatrix} 4 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 27 \\ -4 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} - \begin{bmatrix} 2 \\ 6 \end{bmatrix} + 7 \begin{bmatrix} 4 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 30 \\ 26 \end{bmatrix} = 4 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 3 \begin{bmatrix} 2 \\ 6 \end{bmatrix} + 5 \begin{bmatrix} 4 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 13 \\ 12 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + \begin{bmatrix} 2 \\ 6 \end{bmatrix} + 2 \begin{bmatrix} 4 \\ 0 \end{bmatrix}$$

Express each row matrix of AB as a linear combination of the row matrices of B .

$$AB = \begin{bmatrix} 3 \\ 2 \\ 0 \end{bmatrix} \begin{bmatrix} 6 & -2 & 4 \end{bmatrix} + \begin{bmatrix} -2 \\ 5 \\ 4 \end{bmatrix} \begin{bmatrix} 0 & 1 & 3 \end{bmatrix} + \dots$$

PROOF (b) Show that if A is an $m \times n$ matrix & $A(BA)$ is defined, then B is an $n \times m$ matrix.

Mathematically :

→ Let A is $m \times n$ matrix & B is $m_1 \times n$, matrix

→ If $A(BA)$ is defined then first (BA) should be defined so $n_1 = m$ so (BA) becomes $m_1 \times n$ size matrix

→ Now if $A(BA)$ is defined then $n = m_1$ so the size becomes $m \times n$.

→ So this implies matrix B becomes $n \times m$.

PROOF (a) Show that if A has a row of zeros and B is any matrix for which AB is defined, then AB also has row of zeros.

Solution

Assume that the entries of i -th row of A are all zeros. We claim that the i -th row of AB is a row of zeros.

To see this pick an entry c_{ij} in i -th row of AB . By definition of multiplication of AB , we have.

$$c_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \dots + a_{in}b_{nj} = \sum_{k=1}^n a_{ik}b_{kj}$$

Since the i -th row of A is zero we have $a_{i1} = a_{i2} = \dots = a_{in}$

$$c_{ij} = 0b_{1j} + 0b_{2j} + \dots + 0b_{nj} = \sum_{k=1}^n 0b_{kj} = 0$$

Hence i -th row of AB is a row of zeros.

day / date:

A is
→ If any matrix and C is any scalar, then
the product cA is the matrix obtained by
multiplying each entry of A by c.

If $A = \begin{bmatrix} 2 & 3 & 4 \\ 1 & 3 & 1 \end{bmatrix}$ then $2A = \begin{bmatrix} 4 & 6 & 8 \\ 2 & 6 & 2 \end{bmatrix}$

NOTE: Two matrices A and B can't if $AB = BA$
but in general $AB \neq BA$ but in general
 $AB \neq BA$

NOTE: If A, B and C are matrices such that
 AB and BC are defined then $A(BC) = (AB)C$
which is also called ASSOCIATIVE LAW for
multiplication



KAGHAZ

BASIC RESULTS:

① If A is $m \times n$ matrix and d is diagonal matrix of order m then find the multiplication rule for DA .

SOLUTION DA

$$= \begin{bmatrix} d_1 & 0 & \dots & 0 \\ 0 & d_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \vdots & \dots & d_m \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}$$

$$= \begin{bmatrix} d_1 a_{11} & d_1 a_{12} & \dots & d_1 a_{1n} \\ d_2 a_{21} & d_2 a_{22} & \dots & d_2 a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ d_m a_{m1} & d_m a_{m2} & \dots & d_m a_{mn} \end{bmatrix}$$

∴ DA is obtained by multiplying d_1 with the first row of A , d_2 with second row of A , ... and finally d_m with m th row of A .

② If A is $m \times n$ matrix and $E = \text{diag}(e_1, e_2, \dots, e_n)$ then AE is obtained by multiplying the first column of A by e_1 , ..., n th column of A by e_n .

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} e_1 & \dots & 0 \\ 0 & e_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & \dots & e_n \end{bmatrix}$$



day / date:

$$e_1 a_{11} \ e_2 a_{12} \ \dots \ e_n a_{1n}$$

$$e_1 a_{m1} \ e_2 a_{m2} \ \dots \ e_n a_{mn}$$

- ③ $AE = A$ and $DA = A$, what should be the results values of d_1, d_2, \dots, d_m and e_1, e_2, \dots, e_n

ANSWER only possible values are

$$d_1 = d_2 = \dots = d_m = 1 \text{ and } e_1 = e_2 = \dots = e_n = 1$$

in both cases the diagonal matrices E and D become special matrix which is called

IDENTITY MATRIX.

Notation: $E = I_n$ OR Just I diagonal

$D = I_m$ OR Just I

matrix =
identity
matrix

$$I = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & & \vdots \\ 0 & 0 & \dots & & 1 \end{bmatrix}$$

- ④ If A is a matrix of order n then $A I_n = I_n A = A$, this means that I_n commutes with every matrix with which it multiplies.

INVERSE MATRICES:

→ If Q is square matrix and A is also a square matrix (Both of order n) then Q is said to be an inverse of A if and only if \rightarrow one holds

$$QA = A\bar{Q} = I_n = I$$

is denoted by A^{-1} i.e. $Q = A^{-1}$

* Note: ① The matrix A is described as NONSINGULAR OR INVERTIBLE if an inverse of A exists and singular if no inverse of A exists.

② Both Q and A are square matrices of order n because

they are commuting to give I_n .

→ If B and C are both inverses of matrix A then $\underline{B = C}$

PROOF : $\because B$ is an inverse of A , we have $BA = I \quad \text{--- } ①$

Multiplying both sides (of ①) on the right by C gives

$$(BA)C = IC = C \quad \text{--- } ②$$

* RESULT: Inverse of A but $(BA)C = B(AC) \quad \therefore AC = I$

$$= B(I) = B \quad \text{--- } ③$$

unique.

so that $C = B$ from ② and ③

day / date:

**

→ If A and B are invertible matrices of same size, then $(AB)^{-1} = B^{-1}A^{-1}$

PROOF:

Consider $(AB)(\underline{B^{-1}A^{-1}}) = A(\underline{BB^{-1}})A^{-1}$
 $= A(I)A^{-1}$

$$= AA^{-1} = I - \textcircled{1}$$

also

$$(B^{-1}A^{-1})(AB) = \underline{B^{-1}(A^{-1}A)}B = B^{-1}IB$$

$$= B^{-1}B = I - \textcircled{2}$$

from \textcircled{1} and \textcircled{2}

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

RESULT: hence if A_1, A_2, \dots, A_n are invertible matrices of same size then

$$(A_1 A_2 \dots A_n)^{-1}$$

 $= A_n^{-1} A_{n-1}^{-1} \dots A_2^{-1} A_1^{-1}$

SUMMARY

① A product of any number of invertible matrices is invertible

② The inverse of the product is the product of inverses in the reverse order.



KAGHAZ
www.kaghaz.pk

① If A is an invertible matrix then:

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

PROOF:

$$\because AA^{-1} = A^{-1}A = I$$

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

② $(A^n)^{-1} = (A^{-1})^n$ for $n = 0, 1, 2, \dots$

TRANSPOSE OF A MATRIX

→ if A is $m \times n$ matrix then a transpose of A (denoted by A^T) is obtained by interchanging rows and columns of A .
order (size) of $A^T = n \times m$

Example: $A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix} \rightarrow 2 \times 3$

* first row of A becomes first column of A^T

$$AT = \begin{bmatrix} a_{11} & a_{21} \\ a_{12} & a_{22} \\ a_{13} & a_{23} \end{bmatrix} \rightarrow 3 \times 2$$

$$(A^T)^T = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix}$$

$$(A^T)^T = A$$

**

→ If A and B are of same size then
 $(A+B)^T = A^T + B^T$

Example :

$$\text{let } A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

$$B = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

$$(A+B)^T = \begin{bmatrix} a_{11}+b_{11} & a_{21}+b_{21} \\ a_{21}+b_{21} & a_{22}+b_{22} \end{bmatrix}^T - \textcircled{1}$$

also $A^T + B^T$

$$= \begin{bmatrix} a_{11} & a_{21} \\ a_{12} & a_{22} \end{bmatrix} + \begin{bmatrix} b_{11} & b_{21} \\ b_{12} & b_{22} \end{bmatrix}$$

$$= \begin{bmatrix} a_{11}+b_{11} & a_{21}+b_{21} \\ a_{12}+b_{12} & a_{22}+b_{22} \end{bmatrix} - \textcircled{2}$$

$$\therefore (A+B)^T = A^T + B^T$$

* in transpose left to right diagonal remains same while top right to left gets swapped.

* → A square matrix A is called **SYMMETRIC**
if $A^T = A$

(1) Identity matrix is symmetric

$$\therefore I^T = I \text{ (ALWAYS)}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

(2) every diagonal matrix is symmetric

$$\begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix}^T = \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix}$$

when taking the transpose, diagonal entries of a square matrix don't change their position.

* → A square matrix A is called **SKew-SYMMETRIC**
if $A^T = -A$

example $B = \begin{bmatrix} 0 & 1 & 2 \\ -1 & 0 & 3 \\ -2 & -3 & 0 \end{bmatrix}$ is skew-symmetric

$$\therefore B^T = \begin{bmatrix} 0 & -1 & 2 \\ 1 & 0 & -3 \\ 2 & 3 & 0 \end{bmatrix} = -B$$



day / date:

diagonal entries in skew symmetric matrix
are always zero.

* Theorem says that if A is invertible then
 $Ax = 0$ has trivial solution.

attempt to reduce A to row echelon form

$$\left[\begin{array}{ccc|c} a_{11} & a_{12} & a_{13} & b_1 \\ a_{21} & a_{22} & a_{23} & b_2 \\ a_{31} & a_{32} & a_{33} & b_3 \end{array} \right] \xrightarrow{\text{Row operations}} \left[\begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right]$$

matrix has rank equal to number of non-zero entries

rank of matrix without nonzero entries is

number of zero entries

rank of matrix is not equal to number of non-zero entries

$$A = A^T \Rightarrow A$$

$$\left[\begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right] \xrightarrow{\text{Row operations}} \left[\begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array} \right]$$



KAGHAZ
www.kaghazpk

* → If $AX = B$ represents a system of n equations in n variables then prove that solution is unique if A is invertible.

→ ability to take inverse of matrix

PROOF

~~PROOF~~ let say we have two solutions x_1 and x_2

$$AX_1 = B \quad , \quad AX_2 = B$$

$$AX_1 = AX_2$$

Because A is invertible.

$$A^{-1}AX_1 = A^{-1}AX_2 \quad \therefore A^{-1}A = I$$

$$IX_1 = IX_2$$

$$X_1 = X_2$$

(one solution = unique solution)

∴ $X = A^{-1}B$ is a solution OR

of $AX = B$ (provided $A^{-1}AX = A^{-1}B$

A is invertible) $X = A^{-1}B$ is UNIQUE

How To FIND A^{-1} ?

For this we start elementary row operations.

Elementary row operations are the following :

✓ ① Multiply a row by a nonzero constant.

e.g. $B = \begin{bmatrix} 1 & 0 & 2 & 3 \\ 2 & -1 & 3 & 6 \\ 1 & 4 & 4 & 0 \end{bmatrix}$

R_n : nth row

consider $R_2 \rightarrow 2R_2$, gives

$$A = \begin{bmatrix} 1 & 0 & 2 & 3 \\ 4 & -2 & 6 & 12 \\ 1 & 4 & 4 & 0 \end{bmatrix}$$

✓ ② Interchange two rows

e.g. consider $(R_1 \leftrightarrow R_2)$ gives

$$C = \begin{bmatrix} 2 & -1 & 3 & 6 \\ 1 & 0 & 2 & 3 \\ 1 & 4 & 4 & 0 \end{bmatrix} \text{ from } B$$

✓ ③ Add a multiple of one row to another row

e.g. for $B = \begin{bmatrix} 1 & 0 & 2 & 3 \\ 2 & -1 & 3 & 6 \\ 1 & 4 & 4 & 0 \end{bmatrix}$

consider $R_1 \rightarrow R_1 + 2R_2$ gives

$$D = \begin{bmatrix} -5 & -2 & 8 & 15 \\ 2 & -1 & 3 & 6 \\ 1 & 4 & 4 & 0 \end{bmatrix}$$

EQUIVALENT MATRICES

→ If B is a matrix and A is a matrix obtained from B by one or more elementary row operations then A is called ROW EQUIVALENT to B and vice versa and is denoted by $A \sim B$ or $B \sim A$

ELEMENTARY MATRICES

An $n \times n$ matrix is called an **elementary matrix** if it can be obtained from the $n \times n$ **identity matrix** I_n by performing a **single elementary row operation**.

e.g. $E = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 3 & 0 & 1 \end{bmatrix}$ is an

elementary matrix

$$\therefore \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \sim R_3 \rightarrow \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 3 & 0 & 1 \end{bmatrix} \quad R_3 + 3R_1$$

i.e. E is obtained from I_3

by performing the ERO

$$R_3 \rightarrow R_3 + 3R_1$$

* Apply $R_3 \rightarrow R_3 + 2R_1$ on identity matrix

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix} = E$$

when we perform EA we get B .

* if you apply certain change in A you get B .

day / date:

let $A = \begin{bmatrix} 1 & 0 & 2 \\ 2 & -1 & 3 \\ 1 & 4 & 4 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 2 \\ 2 & -1 & 3 \\ 4 & 4 & 10 \end{bmatrix} = B$

$$R_3 \rightarrow R_3 + 3R_1$$

CONSIDER

$$= \left[\begin{array}{ccc|cc} 1 & 0 & 0 & 1 & 0 & 2 \\ 0 & 1 & 0 & 2 & -1 & 3 \\ 3 & 0 & 1 & 1 & 4 & 4 \end{array} \right]$$

$$= \begin{bmatrix} 1 & 0 & 2 \\ 2 & -1 & 3 \\ 4 & 4 & 10 \end{bmatrix} = B$$

the ERO has the same effect on a matrix as premultiplication of an elementary matrix

or in mathematical language if θ be any ****** ERO and E be the elementary matrix corresponding to θ then

$$\theta(A) = EA = B \quad A \sim B$$

let $\theta_1, \theta_2, \dots, \theta_n$ be n number of EROs and



KAGHAZ
www.kaghaz.pk

E_1, E_2, \dots, E_n are the corresponding elementary matrices such that when applied on an invertible matrix 'A' give 'I' (identity matrix)

i.e. $\theta_1 \dots \theta_2 \theta_n (A) = E_1 \dots E_2 E_1 A = I$, or $PA = I$,

$P = E_1 \dots E_2 E_1 \therefore A$ is invertible

$$\Rightarrow PAA^{-1} = IA^{-1} \Rightarrow PI = A^{-1}$$

now $PA = I \Rightarrow A \sim I$

and $PI = A^{-1} \Rightarrow I \sim A^{-1}$

\therefore EROS which transformed A into I also

transformed I into A^{-1} $\therefore A^{-1}$ can be found by

$$[A; I] \xrightarrow{\text{E.R.O.S}} [I; A^{-1}]$$

METHOD TO FIND A^{-1}

$\rightarrow AX = B$ is called a consistent system of linear equations if there is atleast one solution, otherwise its called inconsistent

\rightarrow If $B \neq 0$ in $AX = B$, then $AX = B$ is called a NON-HOMOGENOUS SYSTEM

\rightarrow if $B = 0$ then $AX = B = 0$ is called a HOMOGENOUS SYSTEM

NOTE:- Homogeneous system is ALWAYS consistent

$\because X = 0$ is always a solution of $AX = 0$ called a zero solution or trivial solution

\rightarrow to be consistent a row has to be completely zeros

~~Q~~: solve the following system (NON-HOMOGENOUS)

of linear equations by finding the inverse of the coefficient matrix.

$$3x_1 + 4x_2 + 5x_3 = 12$$

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

$$2x_1 + x_2 + 3x_3 = 6$$

$$\underline{AX=B}$$

$$\Rightarrow \left[\begin{array}{ccc|c} 3 & 4 & 5 & x_1 \\ 1 & -1 & 2 & x_2 \\ 2 & 1 & 3 & x_3 \end{array} \right] = \left[\begin{array}{c} 12 \\ 2 \\ 6 \end{array} \right]$$

COEFFICIENT
MATRIX

consider

$$\left[A : I \right] = \left[\begin{array}{ccc|cc|cc} 3 & 4 & 5 & 1 & 0 & 0 \\ 1 & -1 & 2 & 0 & 1 & 0 \\ 2 & 1 & 3 & 0 & 0 & 1 \end{array} \right]$$

1st Row 1st Col

STEP 1: Make blue entry $\underline{(1,1)=1}$ by any ERO for this perform $\underline{R_1 \leftrightarrow R_2}$ to get

$$\left[A : I \right] \sim \left[\begin{array}{ccc|cc|cc} 1 & -1 & 2 & 0 & 1 & 0 \\ 3 & 4 & 5 & 1 & 0 & 0 \\ 2 & 1 & 3 & 0 & 0 & 1 \end{array} \right]$$

STEP 2: Make entries $(2,1)$ and $(3,1) = 0$
 for this perform the following two ^{ERo} operations

$$R_2 \rightarrow R_2 - 3R_1$$

$$R_3 \rightarrow R_3 - 2R_1$$

to get

$$\sim \left[\begin{array}{ccc|ccccc} 1 & -1 & 2 & 1 & 0 & 1 & 0 \\ 0 & 7 & -1 & 1 & 1 & -3 & 0 \\ 0 & 3 & -1 & 0 & 0 & -2 & 1 \end{array} \right]$$

STEP 3: Next to make the entry $(2,2) = 1$.
 for this perform $R_2 \rightarrow R_2 - 2R_3$ to

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

STEP 4: Make entries $(1,2)$ and $(3,2) = 0$, for this
 perform

$$R_1 \rightarrow R_1 + R_2$$

$$R_3 \rightarrow R_3 - 3R_2 \text{ to get}$$

$$\sim \left[\begin{array}{ccc|ccccc} 1 & 0 & 2 & 1 & 2 & -2 & 0 \\ 0 & 1 & 1 & 1 & 1 & -2 & 0 \\ 0 & 0 & -4 & -3 & -5 & 7 & 0 \end{array} \right]$$

day / date:

STEP 5: Make entry $(3,3) = 1$ for this perform

$$R_3 \rightarrow \frac{1}{4} R_3 \text{ to get}$$

$$\sim \left[\begin{array}{ccc|ccc} 1 & 0 & 3 & 1 & 2 & -2 \\ 0 & 1 & 1 & 1 & 0 & -2 \\ 0 & 0 & 1 & 3/4 & 5/4 & -7/4 \end{array} \right]$$

STEP 6: In the final step perform $R_1 \rightarrow R_1 - 3R_3$

and $R_2 \rightarrow R_2 - R_3$ to get

$$\sim \left[\begin{array}{ccc|ccc} 1 & 0 & 0 & -5/4 & -7/4 & 13/4 \\ 0 & 1 & 0 & 1/4 & -1/4 & -1/4 \\ 0 & 0 & 1 & 3/4 & 5/4 & -7/4 \end{array} \right]$$

$$A^{-1} = \frac{1}{4} \left[\begin{array}{ccc} -5 & -7 & 13 \\ 1 & -1 & -1 \\ 3 & 5 & -7 \end{array} \right]$$

$$X = A^{-1} B$$

$$X = \frac{1}{4} \left[\begin{array}{ccc} -5 & -7 & 13 \\ 1 & -1 & -1 \\ 3 & 5 & -7 \end{array} \right] \left[\begin{array}{c} 12 \\ 2 \\ 6 \end{array} \right] \quad \underline{\underline{X = A^{-1} B}}$$

$$= \frac{1}{4} \left[\begin{array}{c} -60 - 14 + 78 \\ 12 - 2 - 6 \\ 36 + 10 - 42 \end{array} \right] = \frac{1}{4} \left[\begin{array}{c} 4 \\ 4 \\ 4 \end{array} \right] = \left[\begin{array}{c} 1 \\ 1 \\ 1 \end{array} \right] = \left[\begin{array}{c} x_1 \\ x_2 \\ x_3 \end{array} \right]$$



PROOF

(d) Show that if A is invertible then prove
that $(A^{-1})^T = (A^T)^{-1}$

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

We know that $I = I$

$$(AA^{-1}) = I$$

Apply transpose on both sides

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

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

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

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

$$(A^{-1})^T = (A^T)^{-1} \therefore \text{hence proven}$$

Q10) Consider the matrix $A = \begin{bmatrix} 1 & 0 \\ -5 & 2 \end{bmatrix}$

$$\begin{bmatrix} 1 & 0 \\ 0 & 1/2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 5 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -5 & 2 \end{bmatrix} = I_2$$

$$\begin{bmatrix} 1 & 0 \\ 0 & 1/2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 5 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 5/2 & 1/2 \end{bmatrix}$$

GAUSSIAN ELIMINATION: (Echelon form) A matrix having the following properties is said to be in row-echelon form

(1) If a row does not consist entirely of zeros, then the first nonzero number in the row is a 1. (we call this leading 1)

(2) If there are any rows that consist entirely of zeros, then they are grouped together at the bottom of the matrix.

(3) In any two successive rows that do not consist entirely of zeros, the leading 1 in the lower row occurs farther to the right than the leading 1 in the higher row.

EXAMPLE : Solve the following system by gaussian elimination method

$$3x_1 + 4x_2 + 5x_3 = 12$$

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

$$2x_1 + x_2 + 3x_3 = 6$$

SOLUTION:

$$\left[\begin{array}{ccc|c} 3 & 4 & 5 & 12 \\ 1 & -1 & 2 & 2 \\ 2 & 1 & 3 & 6 \end{array} \right]$$

S STEP 1 Make $(1,1) = 1$, $R_1 \leftrightarrow R_2$

$$\left[\begin{array}{ccc|c} 1 & 1 & 2 & 2 \\ 3 & 4 & 5 & 12 \\ 2 & 1 & 3 & 6 \end{array} \right]$$

STEP 2 Make $(2,1) = 0$ and $(3,1) = 0$,

$$R_2 \rightarrow R_2 - 3R_1, \quad R_3 \rightarrow R_3 - 2R_1$$

$$\left[\begin{array}{ccc|c} 1 & 1 & 2 & 2 \\ 0 & 1 & -1 & 4 \\ 0 & 3 & -1 & 2 \end{array} \right]$$

STEP 3 Make $(2,2) = 1$ $R_2 \rightarrow R_2 - R_3$

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

STEP 4 Make $(3, 2) = 0$, $R_3 \rightarrow R_3 - 3R_2$

$$\left[\begin{array}{ccc|c} 1 & -1 & 2 & 2 \\ 0 & 1 & 1 & 2 \\ 0 & 0 & -4 & -4 \end{array} \right]$$

STEP 5 Make $(3, 3) = 1$, $R_3 \rightarrow \frac{-1}{4}R_3$

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

which is the required echelon form, so the given linear system is reduced to

$$x_1 - x_2 + 2x_3 = 2 \quad \text{--- (1)}$$

$$x_2 + x_3 = 2 \quad \text{--- (2)}$$

$$x_3 = 1 \quad \text{--- (3)}$$

$$x_1 = 1, x_2 = 1, x_3 = 1$$

NOTES

- (1) In row reduction process don't perform any steps by which you lose zeros or 1's
- (2) If possible then avoid the formation of fractions.

A matrix is in reduced row echelon form if

(1) it is already in echelon form

(2) each column that contains a leading 1 has zeros everywhere else.

EXAMPLES:- The following matrices are in the reduced row-echelon form

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

②
$$\begin{bmatrix} 0 & 1 & -2 & 0 & 1 \\ 0 & 0 & 0 & 1 & 3 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

EXAMPLE Solve the following linear system by reducing the augmented matrix to reduced row-echelon form

$$3x_1 + 4x_2 + 5x_3 = 12$$

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

$$2x_1 + x_2 + 3x_3 = 6$$

SOLUTION Here the augmented matrix is

$$\left[\begin{array}{cccc|c} 3 & 4 & 5 & 12 \\ 1 & -1 & 2 & 2 \\ 2 & 1 & 3 & 6 \end{array} \right]$$

and its echelon form is given by

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

(derived last time)

now to reduce it to reduced row-echelon form we proceed as follows:

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

$$\sim \left[\begin{array}{cccc|c} 1 & 0 & 3 & 4 \\ 0 & 1 & 1 & 2 \\ 0 & 0 & 1 & 1 \end{array} \right] \quad R_1 \rightarrow R_1 + R_2$$

$$\sim \left[\begin{array}{cccc|c} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{array} \right] \quad R_1 \rightarrow R_1 - 3R_3, \\ R_2 \rightarrow R_2 - R_3,$$

now rewriting the linear ~~algebra~~ system
again we get

$$x_1 = 1, x_2 = 1, x_3 = 1$$

required reduced row echelon form. In this case entries in the last (4th) column form the solution.

F

FUNDAMENTAL OF LINEAR ALGEBRA

If A is a matrix of order $m \times n$ then
^{statement} there are equivalent

(a) A is invertible

(b) $Ax=0$ has only trivial solution

(c) A can be transformed into row-reduced echelon form which is equal to I_m .

(d) A can be expressed as the product of
any elementary matrices.

$a \Rightarrow b$

PROOF A is $m \times n$ matrix where (more) consider
A is invertible and want to prove
that $Ax=0$ has only trivial solution

Solution

$$AA^{-1}x = A^{-1}0$$

$$\therefore AA^{-1} = I$$

$$Ix = 0$$

$$\therefore Ix = x$$

$$x = 0$$

$b \Rightarrow c$

PROOF

consider $Ax = 0$ has only trivial solution
and want to prove that A can be
transformed into I_n or row-reduced
echelon form.

$$\left[\begin{array}{cccc|c} a_{11} & a_{12} & \dots & a_{1n} & 0 \\ a_{21} & a_{22} & \dots & a_{2n} & 0 \\ a_{n1} & a_{n2} & \dots & a_{nn} & 0 \end{array} \right]$$

By performing EROS A can be transformed

$$\left[\begin{array}{cccc|c} 1 & 0 & \dots & 0 & 0 \\ 0 & 1 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & 0 \\ 0 & 0 & \dots & 1 & 0 \end{array} \right] I_m$$

$c \Rightarrow d$ PROOF

Consider A can be rewritten in the row-reduced echelon form and we want to show that A can be expressed as the product of elementary matrix.

$$\left[A : I \right] \xrightarrow{\text{ERO}} \left[I : A^{-1} \right]$$

$$\begin{array}{ccccc} A & & I & & \\ \downarrow \text{ERO} & & \downarrow \text{ERO} & & \\ B & & E & & \\ & & & EA = B & \end{array}$$

$$E_k E_{k-1} \cdots E_2 E_1 A = I$$

$$PA = I$$

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

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

$$I \sim A^{-1}$$

$$P = E_k^{-1} E_{k-1}^{-1} \cdots E_2^{-1} E_1$$

 $d \Rightarrow a$
PROOF

By using $(AB)^{-1} = A^{-1}B^{-1}$ theorem

QUESTION :

For what values of 'k' does the following system have

(a) NO SOLUTION (b) ONLY ONE SOLUTION

(c) INFINITELY MANY SOLUTIONS.

$$kx + y = 1$$

$$x + ky = 1$$

SOLUTION The augmented matrix of the given system of equation is given by

$$A = \begin{bmatrix} k & 1 & | & 1 \\ 1 & k & | & 1 \end{bmatrix}, \text{ let us try to}$$

find the echelon form of this matrix.

$$\Rightarrow A \sim \begin{bmatrix} 1 & k & | & 1 \\ (k) & 1 & | & 1 \end{bmatrix} \quad R_1 \leftrightarrow R_2$$

→ after making diagonal 1 make

$$\sim \begin{bmatrix} 1 & k & | & 1 \\ (0) & 1 - k^2 & | & 1 - k \end{bmatrix} \quad R_2 \rightarrow R_2 - kR_1$$

Let us discuss different cases;

(1) for exactly one solution $1 - k^2 \neq 0$

① must be transformed into the echelon

day / date:

form by making the entry (2,2) one by performing $R_2 \rightarrow \frac{R_2}{1-k^2}$ to get

$$\sim \left[\begin{array}{ccc|c} 1 & k & 1 & 1 \\ 0 & 1 & 1 & 1+k \end{array} \right]$$

provided

$$\begin{aligned} 1-k^2 &\neq 0 \\ k^2 &\neq 1 \end{aligned}$$

$\Rightarrow k \neq \pm 1 \rightarrow$ for one solution

using $k = -1$ in ① gives

$$\left[\begin{array}{ccc|c} 1 & k & 1 & 1 \\ 0 & 1-k^2 & 1-k & 1-k \end{array} \right] = \left[\begin{array}{ccc|c} 1 & k & 1 & 1 \\ 0 & 0 & 1 & 2 \end{array} \right] \quad \begin{matrix} k = -1 \\ 0 = 2 \text{ (no solution)} \end{matrix}$$

so we have no solution for $k = -1$

second row gives $0 = 2$ which is not possible

using $k = +1$ in ① gives.

$$\left[\begin{array}{ccc|c} 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 \end{array} \right], \text{ rewriting the linear } \\ 0=0 \text{ (infinitely many solutions)}$$

system gives $x_1 + x_2 = 1$

here no of unknowns = 2, no of equations = 1



no of equations < no of unknowns
→ infinite solutions. day/date:

which is less than no of unknowns. This gives
infinite solutions for $k=+1$.

NOTE: Here (infinite solutions case)

x_1 is called a leading variable which corresponds to the leading 1 in the echelon form

i.e. $\left[\begin{array}{ccc|c} 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 \end{array} \right]$ and x_2 is

does not

call a free variable which correspond to the leading 1!

We can write down solution like

$$x_1 = 1 - x_2 = 1 - t \quad \text{where } x_2 = t$$

we have infinite values for x_1 and x_2

∴ we have infinite solutions for $k=1$.

Question 16

Find the condition that the b's must satisfy for the system to be consistent

$$6x_1 - 4x_2 = b_1$$

$$3x_1 - 2x_2 = b_2$$

OR solution:

$$\begin{array}{l} R_1 \rightarrow R_1 - 2R_2 \\ \left[\begin{array}{cc|c} 6 & -4 & b_1 \\ 3 & -2 & b_2 \end{array} \right] \xrightarrow{R_2 \leftrightarrow 2R_1} \left[\begin{array}{cc|c} 3 & -2 & b_2 \\ 6 & -4 & b_1 \end{array} \right] \end{array}$$

$$b_1 - 2b_2 = 0$$

$$2b_2 = b_1$$

$$\xrightarrow{R_2 - 2R_1} \left[\begin{array}{ccc|c} 3 & -2 & 1 & b_2 \\ 0 & 0 & 1 & b_1 - 2b_2 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

one can see that, to be consistent
second row has to get completely zeros.

$$b_1 = 2b_2$$

Question 23

since $Ax = 0$ has only $x = 0$ as a solution

Theorem 1.6.4 guarantees that $A \neq 0$ is invertible.

By Theorem 1.4.8 (b), A^k is also invertible.

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

$$\underbrace{A^{-1} A^{-1} \dots A^{-1}}_{k \text{ factors}} \underbrace{AA \dots A}_{n \text{ factors}} = I$$

Because A^k is invertible, Theorem 1.6.4 allows us to conclude that $A^k x = 0$ has only trivial solution.



$$A^k x = 0$$

$$(A^k)^{-1} A^k x = (A^k)^{-1} 0$$

$$\underbrace{A^{-1} A^{-1} \dots A^{-1}}_{k \text{ factors}} \underbrace{AA \dots A}_{n \text{ factors}} x = 0$$

$$(I)x = 0 \quad \because IA = A$$

$x = 0$ has only trivial sol.

Question 24

Let $Ax=0$ be a homogeneous solution of n linear eqs in n unknown and let

Q be an invertible $n \times n$ matrix.

Show that $Ax=0$ has just the trivial solution if and only if $(QA)x=0$ has just trivial solution.

PROOF

Let $Ax=0$ holds. Now we apply Q matrix from LHS we will have

$$Q(Ax) = Q0$$

$(QA)x = (0 + \dots)$: Associative property.

Now we let $(QA)x=0$ and we apply Q^{-1}

$$Q^{-1}QAx = Q^{-1}0$$

$$IAx = 0$$

$$Ax = 0$$

Question 25

Suppose that x_1 is a fixed matrix which satisfies the equation $Ax_1 = b$. Further let x be any matrix whatever which satisfies $Ax = b$. We must then show that there is a matrix x_0 which satisfies both eq $x = x_1 + x_0$.

day / date:

and $Ax_0 = 0$. Clearly the ① eq implies that
 $x_0 = x_1 - x$

Given

$$x = x_1 + x_0$$

$$x_0 = x_1 - x$$

$$A(x_0) = A(x_1 - x)$$

$$A(x_0) = A(x_1) - A(x)$$

$$= b - b = 0$$

$$A(x_0) = 0$$

Now we show that $A(x_1 + x_0)$ has solution

$$A(x_1 + x_0) = A(x_1) + A(x_0)$$

$$A(x_1 + x_0) = b + 0$$

$$A(x_1 + x_0) = b$$



KAGHAZ
www.kaghaz.pk

REVISION:-

If A is an invertible square matrix then A^{-1} could be evaluated by the following method:

$$\left[A : I \right] \xrightarrow{\text{EROS}} \left[I : A^{-1} \right]$$

\hookrightarrow identity matrix EROS \rightarrow elementary row operations.

~~PROOF~~

Example

Let $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ then using the technique

$$\left[A : I \right] \xrightarrow{\text{EROS}} \left[I : A^{-1} \right] \text{ prove that}$$

$$A^{-1} = \frac{1}{(ad-bc)} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}, \quad ad-bc \neq 0$$

Solution

$$\left[\begin{array}{cc|cc} a & b & 1 & 0 \\ c & d & 0 & 1 \end{array} \right] \sim \left[\begin{array}{cc|cc} 1 & b/a & 1/a & 0 \\ c & d & 0 & 1 \end{array} \right] \xrightarrow{R_1 \rightarrow R_1/a}$$

$$\sim \left[\begin{array}{cc|cc} 1 & b/a & 1/a & 0 \\ 0 & d - cb/a & -c/a & 1 \end{array} \right] \xrightarrow{R_2 \rightarrow R_2 - CR_1} \left(\frac{ad-bc}{a} \right)$$

$$= \left[\begin{array}{cc|cc} 1 & b/a & 1/a & 0 \\ 0 & \frac{ad-bc}{a} & -\frac{c}{a} & 1 \end{array} \right]$$



day / date:

$$\sim \left[\begin{array}{cc|cc} 1 & b/a & 1/a & 0 \\ 0 & 1 & 1 & \frac{-c}{ad-bc} \\ \end{array} \right]$$

$\hookrightarrow R_2 \rightarrow R_2 - \frac{b}{a} R_1$, $ad-bc \neq 0$

$$\sim \left[\begin{array}{cc|cc} 1 & 0 & \frac{d}{ad-bc} & \frac{-b}{ad-bc} \\ 0 & 1 & 1 & \frac{-c}{ad-bc} \\ \end{array} \right]$$

$\hookrightarrow R_1 \rightarrow R_1 - \frac{b}{a} R_2$

which completes the proof.

RESULT

For $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$

$$A^{-1} = \frac{1}{(ad-bc)} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

provided $ad-bc \neq 0$

this number $ad-bc$ is called determinant of matrix A and is written

DETERMINANT OF 3×3 MATRIX

$$A = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$$

$$|A| = a(ei-fh) + b(di-gf) + c(dh-eg)$$

$$\begin{bmatrix} ax & bf & ce \\ ie & df & bg \\ gh & hi & di \end{bmatrix} - \begin{bmatrix} bx & af & cd \\ fd & eg & ah \\ gi & hi & ei \end{bmatrix} + \begin{bmatrix} cx & bg & ae \\ de & hg & fi \\ jd & ih & gi \end{bmatrix}$$

$$ad-bc = \det(A) = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = |A|$$



here A is invertible.

$\therefore A$ is a square matrix so we have the following result :

determinant of a nonsingular nonsquare matrix is not defined also if A^{-1} exists then $\det(A) \neq 0$

RESULT: if $AX = B$ is the system of n linear equations in n unknowns then :

① Unique solution if A is invertible ($\det(A) \neq 0$) and $X = A^{-1}B$

② If A^{-1} doesn't exist

(a) \downarrow
NO SOLUTION

\downarrow (b)
INFINITE
SOLUTIONS

Let us consider examples when $AX = B$ is the system of two linear equations in two unknowns.

① For unique solution consider the following system $(X = A^{-1}B)$

Consider

$$3x_1 + 4x_2 = 7 \quad \text{--- (1)}$$

$$x_1 - x_2 = 0 \quad \text{--- (2)}$$

* taking determinant
of coefficient matrix

$$\Rightarrow \begin{bmatrix} 3 & 4 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 7 \\ 0 \end{bmatrix}$$

A ↪

B ↪

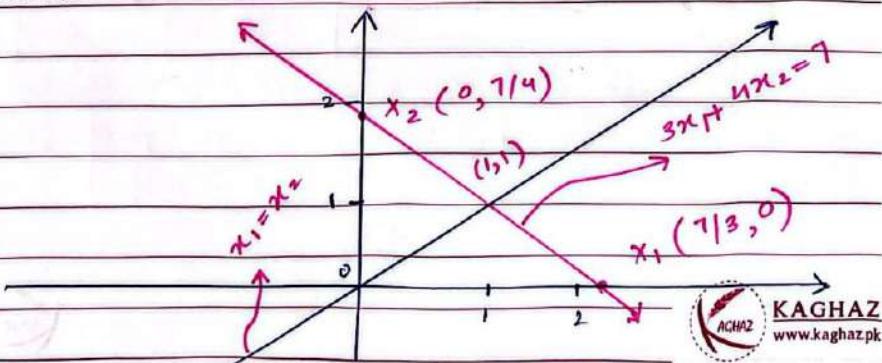
$$\text{now } \begin{vmatrix} 3 & 4 \\ 1 & -1 \end{vmatrix} = -3 - 4 = -7 \neq 0 \Rightarrow |A| \neq 0$$

$$\therefore \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3 & 4 \\ 1 & -1 \end{bmatrix}^{-1} \begin{bmatrix} 7 \\ 0 \end{bmatrix} \quad (X = A^{-1}B)$$

$$= \frac{-1}{7} \begin{bmatrix} -1 & -4 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 7 \\ 0 \end{bmatrix} = \begin{bmatrix} -1 & -4 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 1 & 4 \\ 1 & -3 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \text{ie point of unique solution}$$

intersection of lines (1) and (2) is $(1, 1)$ as shown below.



2a) consider $x_1 + x_2 = 4$
 $x_1 + x_2 = 1 \rightarrow 4 \neq 1$

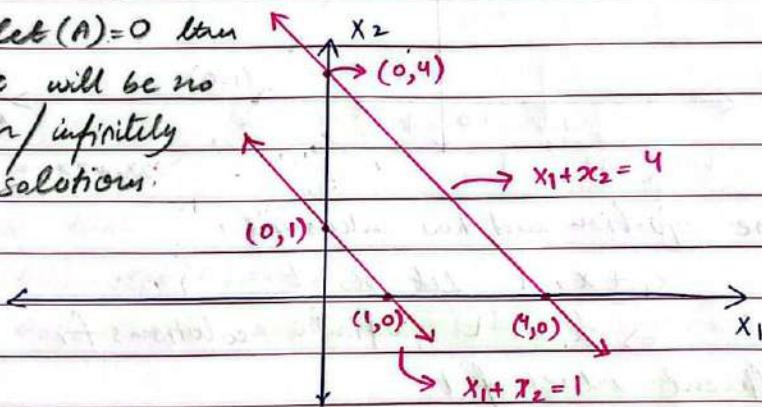
$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 4 \\ 1 \end{bmatrix} \quad (\text{NO SOLUTION})$$

A x B

$$|A| = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = 1 - 1 = 0 \quad \text{ie parallel lines don't intersect.}$$

when $\det(A) = 0$ then

there will be no solution / infinitely many solutions.



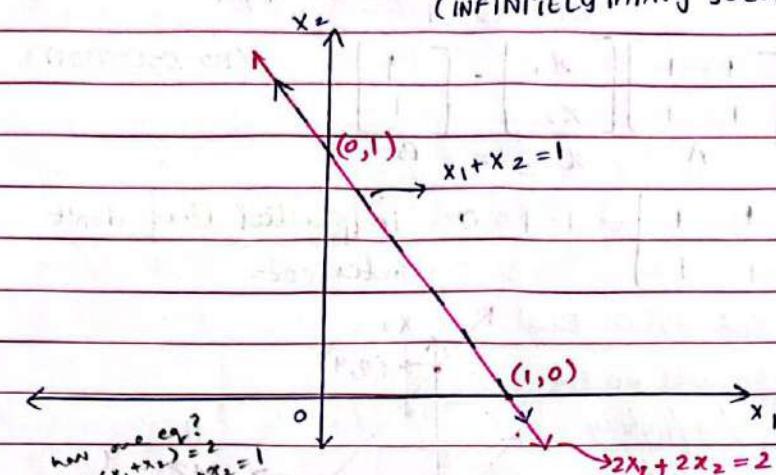
2b) $x_1 + x_2 = 1, 2x_1 + 2x_2 = 2$

$$\begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \quad (\text{INFINITE MANY SOLUTIONS})$$

$$\begin{vmatrix} 1 & 1 \\ 2 & 2 \end{vmatrix} = 2 - 2 = 0$$

both lines are **COINCIDENT**

(INFINITELY MANY SOLUTIONS)



one equation and two unknowns

$$x_1 + x_2 = 1, \text{ let } x_2 = t$$

$\Rightarrow x_1 = 1 - t$, infinite solutions for different values of t .

NOTE :-

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

$$\sim \left[\begin{array}{ccc|cc} 1 & 1 & | & 1 & 0 \\ 0 & 0 & | & -2 & 1 \end{array} \right] R_2 \rightarrow R_2 - 2R_1$$

$$\not\sim \left[\begin{array}{cc|c} 1 & 0 & | & A^{-1} \\ 0 & 1 & | & \end{array} \right], A^{-1} \text{ doesn't exist}$$

→ if A is invertible then matrix A is expressible as product of elementary matrices.

$$A = E_1 E_2 \dots E_r \text{ so}$$

$$AB = E_1 E_2 \dots E_r B$$

$$\det(AB) = \det(E_1) \det(E_2) \dots \det(E_r) \wedge$$

$$\det(AB) = \underbrace{\det(E_1 E_2 \dots E_r)}_{\det(A)} \det(B)$$

$$\det(AB) = \det(A) \det(B)$$

PROPERTIES OF DETERMINANT

① If A and B are square same size then $\det(AB) = \det(E_1 E_2 \dots E_r B) = \det(A) \det(B)$

PROOF 1

CHECK FOR 2×2 MATRICES

for $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$, $\det(A) = a_{11}a_{22} - a_{12}a_{21}$

$$B = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}, \det(B) = b_{11}b_{22} - b_{12}b_{21}$$

PROOF 2

prove that

$$\det(AB) = \det(A) \det(B)$$

$$= a_{11}a_{22}b_{11}b_{22} - a_{11}a_{22}b_{12}b_{21} - a_{12}a_{21}b_{11}b_{22} + a_{12}a_{21}b_{12}b_{21}$$

② If $I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $\det(I) = 1$ which is true for identity matrix of any order.

PROVE THE FOLLOWING.

If A is square matrix A is invertible then

$$\det(A) \neq 0$$

→ if A is NOT invertible then neither is the product AB.

$$\hookrightarrow \det(AB) = 0$$

$$\text{and } \det(A) = 0 \text{ so it}$$

follows that

$$\det(AB) = \det(A) \det(B)$$

det(A) ≠ 0

RESULT If A^{-1} exists then

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

$$\det(A^{-1})\det(A) = 1 = \det(A^{-1}) = 1$$

$\det(A)$

$$\det(A^{-1}) = \frac{1}{\det(A)}$$

$$\det(AB) = \det(A)\det(B)$$

③ $\det(A) = \det(A^T)$

for 2×2 , $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$

$$A^T = \begin{bmatrix} a_{11} & a_{21} \\ a_{12} & a_{22} \end{bmatrix}$$

$$\det(A) = \det(A^T) = a_{11}a_{22} - a_{21}a_{12}$$

Two

④ If two columns or two rows of a matrix are identical then $\det(A) = 0$

Check for 2×2 matrix

$$\begin{vmatrix} a & a \\ b & b \end{vmatrix} = ab - ba = 0$$

OR

$$\begin{vmatrix} a & b \\ a & b \end{vmatrix} = ab - ba = 0$$



day / date:

(5) Adding rows (or columns) together
makes no difference to the determinant

$$\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{12}a_{21}$$

consider $(C_1 \rightarrow C_1 + C_2)$

$$\begin{vmatrix} a_{11} + a_{12} & a_{12} \\ a_{21} + a_{22} & a_{22} \end{vmatrix}$$

$$\Rightarrow a_{11}a_{22} - a_{12}a_{21} \quad (\text{still SAME})$$

$$= \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix}$$

do

$$\begin{array}{c|cc} R_1 \rightarrow R_1 + R_2 & a_{11} + a_{21} & a_{12} + a_{22} \\ \hline & a_{21} & a_{22} \end{array}$$

$$\Rightarrow a_{11}a_{22} - a_{12}a_{21}$$

$$= \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix}$$

$$\det(kA) = k^n \det(A)$$

∴ where n is the number
of rows.

(6) If B is the matrix that results when a single row or single column of A is multiplied by a scalar k , then $\det(B) = k \det(A)$

$$\begin{aligned} \text{Let } \det(B) &= \begin{vmatrix} ka_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = \begin{vmatrix} a_{11} & ka_{12} \\ a_{21} & ka_{22} \end{vmatrix} \\ &= \begin{vmatrix} ka_{11} & ka_{12} \\ a_{21} & a_{22} \end{vmatrix} = \begin{vmatrix} a_{11} & a_{12} \\ ka_{21} & ka_{22} \end{vmatrix} \end{aligned}$$

$$= k(a_{11}a_{22} - a_{21}a_{12}) = k \begin{vmatrix} a_{11} & a_{12} \end{vmatrix}$$

$$\boxed{\det(B) = k \det(A)}$$

TRY THE FOLLOWING

Find $\det(A)$ where

$$A = \begin{bmatrix} b+c & c+a \\ a & b \\ 1 & 1 \end{bmatrix}$$

$$R_1 \rightarrow R_1 + R_2$$

now taking $a+b+c$ common - we get

$$\begin{bmatrix} a+b+c & a+b+c & a+b+c \\ a & b & c \\ 1 & 1 & 1 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 1 & 1 & 1 \\ a & b & c \\ 1 & 1 & 1 \end{bmatrix} a+b+c.$$

→ A is NOT INVERTIBLE thus $\det(A)=0$
 → if two columns/rows are IDENTICAL, then $\det(A)=0$
 → if a square matrix with two proportional rows/columns, then $\det(A)=0$
 → if A has row/column of zeros, then $\det(A)=0$
 → if a scalar k is multiplied $\det(B)=k \det(A)$
 → Adding rows/column, the determinant remain SAME.

* since two rows are identical we get $\det(A)=0$



(7) If A is a square matrix with two proportional rows or two proportional columns, then $\det(A) = 0$

$$\rightarrow c_2 = 2c_3$$

EXAMPLE :- $\det \begin{pmatrix} -2 & 8 & 4 \\ 3 & 2 & 1 \\ 1 & 10 & 5 \end{pmatrix} = 0$

$$\therefore \begin{vmatrix} -2 & 8 & 4 \\ 3 & 2 & 1 \\ 1 & 10 & 5 \end{vmatrix} = \begin{vmatrix} -2 & 4 & 4 \\ 3 & 1 & 1 \\ 1 & 5 & 5 \end{vmatrix} = 0$$

* In determinant the constant 'k' not necessarily needs to be multiplied by each row / col. $\rightarrow c_2$ is identical

(8) If A is a square matrix such that A has a row of zeros or a column of zeros then $\det(A) = 0$

$$\begin{vmatrix} a & 0 \\ b & 0 \end{vmatrix} = \begin{vmatrix} a & b \\ 0 & 0 \end{vmatrix} = a(0) - b(0) = 0$$

NOTE :- If A is any square matrix that contains a row of zeros or a column of zeros then A is singular.



MORE PROPERTIES

① If A and B are square matrices of same size then $\det(A) + \det(B) \neq \det(A+B)$

PROOF

EXAMPLE For $A = \begin{bmatrix} 1 & 2 \\ 2 & 5 \end{bmatrix}$, $B = \begin{bmatrix} 3 & 1 \\ 1 & 3 \end{bmatrix}$, $A+B = \begin{bmatrix} 4 & 3 \\ 3 & 8 \end{bmatrix}$

$$\det(A) = 1, \det(B) = 8$$

$$\det(A+B) = 23$$

$9 \neq 23$ hence proved

TRY THE FOLLOWING

$$\text{Let } A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, B = \begin{bmatrix} a_{11} & a_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

find $\det(A) + \det(B) = ?$

$$\det(A) = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{12}a_{21}$$

$$\det(B) = \det \begin{bmatrix} a_{11} & a_{12} \\ b_{21} & b_{22} \end{bmatrix} \Rightarrow a_{11}b_{22} - a_{12}b_{21}$$

$$\underline{\text{ANS}} \quad a_{11}(a_{22} + b_{22}) - a_{12}(a_{21} + b_{21}) \Rightarrow \det(A)\det(B)$$

which can be written as.

$$\left| \begin{array}{cc} a_{11} & a_{12} \\ a_{21} + b_{21} & a_{22} + b_{22} \end{array} \right| \therefore \text{we have}$$

$$\left| \begin{array}{cc} a_{11} & a_{12} \\ a_{21} & a_{22} \end{array} \right| + \left| \begin{array}{cc} a_{11} & a_{12} \\ b_{21} & b_{22} \end{array} \right|$$

$$= \left| \begin{array}{cc} a_{11} & a_{12} \\ a_{21} + b_{21} & a_{22} + b_{22} \end{array} \right| \det(A+B)$$

- (2) (ADDITION RULE) Let A, B and C be $n \times n$ matrices that differ only in single row (say n^{th}) and assume that r^{th} row of C can be obtained by adding corresponding entries in r^{th} rows of A and B . Then

$$\det(C) = \det(A) + \det(B)$$

- (3) If B is the matrix that results when two rows or two columns of A are interchanged, then

$$\det(B) = -\det(A)$$

~~PROOF~~CHECK for 2×2 matrix

$$\text{If } A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, B = \begin{bmatrix} a_{21} & a_{22} \\ a_{11} & a_{12} \end{bmatrix}$$

$$\begin{aligned}\det(A) &= a_{11}a_{22} - a_{12}a_{21} \\ &= - (a_{12}a_{21} - a_{11}a_{22}) \\ &= - \det(B)\end{aligned}$$

$$\boxed{\det(A) = -\det(B)}$$

TRY THE FOLLOWING

$$\text{If } A = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}, \text{ let } \det(A) = -7$$

$$\text{find } \det \begin{bmatrix} a & g & d \\ b & h & e \\ c & i & f \end{bmatrix} = \begin{vmatrix} a & g & d \\ b & h & e \\ c & i & f \end{vmatrix} = ?$$

SOLUTION

$$\begin{vmatrix} a & g & d \\ b & h & e \\ c & i & f \end{vmatrix} = \begin{vmatrix} a & b & c \\ g & h & i \\ d & e & f \end{vmatrix} \quad \text{taking transpose.}$$

$$= - \begin{vmatrix} a & b & 0 \\ d & e & f \\ g & h & l \end{vmatrix} = - \det(A) = -(-7) = 7$$

 $R_1 \leftrightarrow R_3$

- (4) If B is the matrix that results when A multiple of one row of A is added to another row of or when a multiple of one column is added to another column, then $\det(A) = \det(B) \rightarrow (*)$

CHECK for 2×2 case.

$$\text{Take } A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, B = \begin{bmatrix} a_{11} + ka_{12} & a_{12} \\ a_{21} + ka_{22} & a_{22} \end{bmatrix}$$

$$= \begin{bmatrix} a_{11} + ka_{12} & a_{12} \\ a_{21} + ka_{22} & a_{22} \end{bmatrix}, k \text{ is any scalar } k \neq 0$$

$\hookrightarrow C_1 \rightarrow C_1 + kC_2$

DETAIL

$$\det(B) = \begin{vmatrix} a_{11} + ka_{12} & a_{12} \\ a_{21} + ka_{22} & a_{22} \end{vmatrix} = (a_{11} + ka_{12})a_{22} - a_{12}(a_{21} + ka_{22})$$

$$= a_{11}a_{22} - a_{12}a_{21}$$

$$= \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = \det(A)$$

- (5) If any matrix A is singular then $\det(A) = 0$

CRAMER'S RULE

We shall discuss another method to find the unique solution of n equations in n unknowns, provided determinant of the coefficient matrix $\neq 0$ in $AX=B$ $B \neq 0$ i.e linear system is NON HOMOGENOUS.

SIMPLE CASE

system of two equations in two unknowns:

$$\text{here } AX=B \Rightarrow B = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

$$a_{11}x_1 + a_{12}x_2 = b_1, \quad \text{--- (1)}$$

$$a_{21}x_1 + a_{22}x_2 = b_2 \quad \text{--- (2)}$$

from (1)

$$x_2 = b_1 - a_{11}x_1 \quad \text{--- (*)}$$

a_{12}

using (*) in (2) gives the following

$$x_1 = \frac{b_1 a_{22} - b_2 a_{12}}{a_{11} a_{22} - a_{21} a_{12}}$$

$$x_1 = \frac{\begin{vmatrix} b_1 & a_{12} \\ b_2 & a_{22} \end{vmatrix}}{\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix}}, \text{ similarly we can obtain}$$

day / date:

$$x_2 = \frac{\begin{vmatrix} a_{11} & b_1 \\ a_{21} & b_2 \end{vmatrix}}{\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix}} \text{ or just } \leftarrow$$

$$x_1 = \frac{|A_1|}{|A|}, \quad x_2 = \frac{|A_2|}{|A|}, \text{ where } |A| \text{ is}$$

the different determinant of coefficient matrix

$$|A_1| = \begin{vmatrix} b_1 & a_{12} \\ b_2 & a_{22} \end{vmatrix} \text{ is obtained by replacing}$$

the entries in the 1st column of $|A|$ by the entries in B , etc. Similarly we can extend this method to three equations in three unknowns.

But first we must know the expansion or method to expand a determinant of a matrix of order 3.

EXPANSION BY FIRST ROW:

Consider

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$



$$= a_{11} (-1)^{1+1} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} + a_{12} (-1)^{1+2} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix}$$

$$+ a_{13} (-1)^{1+3} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

$$= a_{11}(a_{22}a_{33} - a_{23}a_{32}) - a_{12}(a_{21}a_{33} - a_{23}a_{31}) \\ + a_{13}(a_{21}a_{32} - a_{22}a_{31})$$

If A is a square matrix then the minor of entry a_{ij} is denoted by M_{ij} and is defined to be the determinant that remains after i^{th} row and j^{th} column are deleted from A the number $(-1)^{i+j} M_{ij}$ is denoted C_{ij} and is called the cofactor of entry a_{ij} .

EXAMPLE

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & \begin{array}{|c c|} a_{22} & a_{23} \\ a_{32} & a_{33} \end{array} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$M_{11} \text{ minor of } a_{11} = \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix}$$

C_{11} = cofactor of a_{11}

$$= (-1)^{1+1} M_{11} = \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix}$$



day / date:

$$M_{12} = \text{minor of } a_{12} = \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} = ad - bc$$

$$\Rightarrow C_{12} = (-1)^{1+2} M_{12} = - \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix}$$

similarly

$$C_{13} = (-1)^{1+3} M_{13} = \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

NOTE: $\det(A) = \det(A)$

$$= a_{11} (-1)^{1+1} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} + a_{12} (-1)^{1+2} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix}$$
$$+ a_{13} (-1)^{1+3} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

$$= a_{11} C_{11} + a_{12} C_{12} + a_{13} C_{13}$$

TRY THE FOLLOWING

Find $C_{11}, C_{12}, C_{21}, C_{22}$ for $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$

$$\text{Solution: } C_{11} = (-1)^{1+1} d = d$$

$$C_{12} = (-1)^{1+2} c = -c$$

$$C_{21} = (-1)^{1+2} b = -b$$

$$C_{22} = (-1)^{2+2} a = a.$$



day / date:

consider $\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}^T = \begin{bmatrix} d & -c \\ -b & a \end{bmatrix}^T$

$$= \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}, \text{ recall that } \quad ①$$

$$\Rightarrow A^{-1} = \frac{1}{(ad-bc)} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \rightarrow ②$$

$$\Rightarrow A^{-1} = \frac{1}{\det(A)} \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}^T$$

from ① and ②
this matrix is
called **Adj(A)** or
Adjoint of A.

INVERSE OF MATRIX USING ITS ADJOINT

If A is an invertible matrix then

$$A^{-1} = \frac{1}{\det(A)} \text{adj}(A)$$

Example: Find the inverse of $A = \begin{bmatrix} 3 & 2 & -1 \\ 1 & 6 & 3 \\ 2 & -4 & 0 \end{bmatrix}$ by adjoint method



KAGHAZ

www.kaghaz.pk

STEP 1 Find $\det(A)$ which is given by

$$3(0+12) - 2(-6) - 1(-4-12) = 36 + 12 + 16 = 64$$

$$\det(A) \neq 0$$

STEP 2 find $\text{adj}(A)$ which is

$$= \begin{bmatrix} C_{11} & C_{12} & C_{13} \\ C_{21} & C_{22} & C_{23} \\ C_{31} & C_{32} & C_{33} \end{bmatrix}^T = \text{adj}(A)$$

** after taking transpose.*

C_{11} = cofactor of $(a_{11} = 3)$ is given by

$$C_{11} = (-1)^{1+1} M_{11} = (-1)^2 \begin{vmatrix} 6 & 3 \\ -4 & 0 \end{vmatrix} = 12$$

$$\text{Similarly } C_{12} = (-1)^{1+2} M_{12} = - \begin{vmatrix} 1 & 3 \\ 2 & 0 \end{vmatrix} = 6$$

check the following

$$C_{13} = \begin{vmatrix} 1 & 6 \\ 2 & -4 \end{vmatrix} = -4 - 12 = -16$$

$$\therefore A^{-1} = \frac{1}{\det(A)} \text{adj}(A) = \frac{1}{64} \begin{bmatrix} 12 & 4 & 12 \\ 6 & 2 & -10 \\ -16 & 16 & 16 \end{bmatrix}$$

PROOF

Question 16 Let A and B be $n \times n$ matrices.

Show that if A is invertible, then $\det(B) = \det(A^{-1}BA)$

solution:

$$\begin{aligned}\det(A^{-1}BA) &= \det(A^{-1}) \det(B) \det(A) \\ &= \frac{1}{\det(A)} \det(B) \det(A)\end{aligned}$$

$$\det(A^{-1}BA) = \det(B) \quad (\text{proven})$$



KAGHAZ
www.kaghaz.pk

PROOF

day / date:

Question 21 Let A and B be $n \times n$ matrices.

You know from earlier work that AB is invertible if A and B are invertible.

What can you say about the invertibility of AB if one or both of the factors are singular? Explain your reasoning.

Solution If either A or B is singular, then either $\det(A) = 0$ or $\det(B) = 0$.

Hence $\det(AB) = \det(A)\det(B) = 0$. Thus AB is also singular.

PROOF

Question 25

Prove that if $\det(A)=1$ and all entries in A are integers, then all the entries in A^{-1} are integers.

Solution This follows from Theorem 2.1.2 and the fact that the cofactors of A are integers if A has only integer entries since integers are closed under multiplication addition & subtraction. $A^{-1} = \frac{1}{\det(A)} \text{adj}(A) = \frac{\text{adj}(A)}{\det(A)}$



KAGHAZ

www.kaghaz.pk

PROOF

If B is $n \times n$ matrix and E is an elementary matrix of order $n \times n$ then
 $\det(EB) = \det(E)\det(B)$

If E is performed by multiplying a row of I_n by scalar of k then this means that

$$EB = C \text{ (say)}$$

$$\det(EB) = \det(C)$$

$$\det(EB) = k \det(B)$$

$$\text{since } \det(E) = k \det(I) = k$$

$$\det(EB) = \det(E)\det(B)$$



KAGHAZ
www.kaghaz.pk

TRIANGULAR MATRICES

(1) A square matrix in which all the entries above the main diagonal are zero is called lower triangle.

Example:

$$A = \begin{bmatrix} a_{11} & 0 & 0 \\ a_{21} & a_{22} & 0 \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

is lower triangle here $\det(A) = a_{11}a_{22}a_{33}$

i.e. the product of diagonal entries which is true for any lower triangular matrix.

(2) Similarly a square matrix in which all the entries below the main diagonal are zero is called UPPER TRIANGULAR MATRIX.

Example:

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} & a_{23} \\ 0 & 0 & a_{33} \end{bmatrix}$$

is upper triangular matrix.

here $\det(A) = a_{11}a_{22}a_{33}$ i.e. product of diagonal entries.

RESULT: A matrix that is either upper or lower triangular is called triangular and

day / date:

determinant of any triangular matrix is equal to product of it's diagonal entries.

NOTE: A square matrix in row echelon form is upper triangular since it has zeros below the main diagonal.

$$I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 4 & -3 \\ 0 & 1 & 6 \\ 0 & 0 & 1 \end{bmatrix} \text{ and } \begin{bmatrix} 0 & 1 & 2 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

are in row-echelon form and also upper triangular matrices.

NOTE: Diagonal matrices are both upper triangular and lower triangular
e.g. $I \rightarrow$ Identity matrix etc.

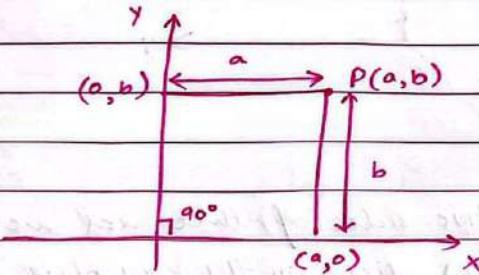


VECTORS

day / date:

2-DIMENSIONAL SPACE

let us consider xy -plane in which any point P is denoted by two numbers, uniquely associated with P , called coordinates e.g. if P is (a, b) then a is the x -coordinate which gives distance from y -axis and b is the y -coordinate which gives distance from x -axis as shown below.



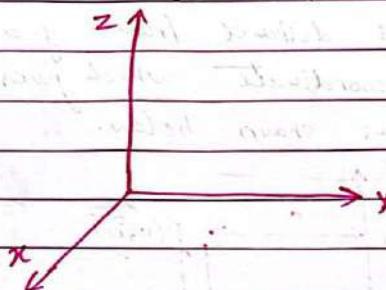
along x -axis, $y=0$ along y -axis, $x=0$.

Now let us consider 3 dimensional space we begin with a set of three lines, called axes concurred at a point (origin)

The three lines (axes) are : (3-dimensional)

- ① NOT COPLANAR i.e. they not all lie in same plane.

- (2) MUTUALLY PERPENDICULAR ie angle b/w them is 90°
- (3) Labeled x, y and z determine a set of three numbers called coordinates, consider a point $P(x, y, z)$ in space below

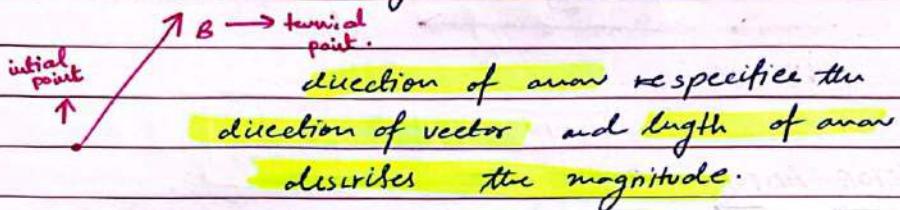


- (4) Only positive sides of three axes are shown.
- (5) $\&(x, y, 0)$ lies in the xy -plane which is formed due to the intersection of x and y axes.
- (6) In xy -plane z coordinate = 0 because z coord gives distance from xy plane
- (7) yz plane is determined from intersection of y and z axes and x coord is 0.
- (8) xz plane is determined due to intersection of x and z axes.

day / date:

VECTORS: Quantities which ~~are~~ are completely determined by magnitude and direction
e.g. displacement, velocity etc.

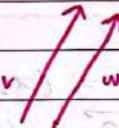
→ geometrically they can be represented as directed line segments or arrows.



Equal (Equivalent Vectors)

Vectors with same length and magnitude.

$$v = v = w$$



Absolute value of scalar.

$$|k| \geq 0, \quad |k| = k, \quad k \geq 0$$

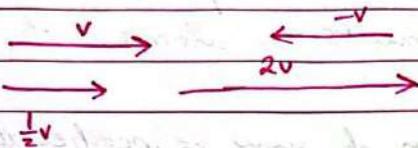
$$|k| = -k, \quad k < 0$$

$$|2| = 2, \quad |-2| = -(-2) = 2.$$

$k \rightarrow$ scalar (real number)

If k is a NONZERO scalar and v is the nonzero vector then kv is the vector whose length is $|kv|$

times the length of v and whose direction is the same as that of v if $k > 0$ and opposite to that of v if $k < 0$.



VECTOR ADDITION :-

$$\overrightarrow{OP_2} = \overrightarrow{OP_1} + \overrightarrow{P_1 P_2}$$

$$\overrightarrow{P_1 P_2} = \overrightarrow{OP_2} - \overrightarrow{OP_1}$$

$$= (x_2 - 0, y_2 - 0, z_2 - 0) - (x_1, y_1, z_1)$$

$$\overrightarrow{P_1 P_2} = (x_2 - x_1, y_2 - y_1, z_2 - z_1)$$

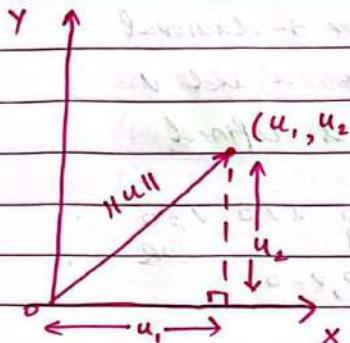
The components of $\overrightarrow{P_1 P_2}$ are obtained by subtracting the coordinates of the initial point from the coordinates of the terminal point.

The length of a vector v is often called the norm of v and is denoted by $\|v\|$ in 2 dimensional space for $v = (u_1, u_2)$

$$\|v\| = \sqrt{u_1^2 + u_2^2} \quad (\text{PYTHAGORAS THEOREM})$$

$$\text{UNIT VECTOR} \rightarrow \frac{\mathbf{v}}{\|\mathbf{v}\|}$$

day / date:

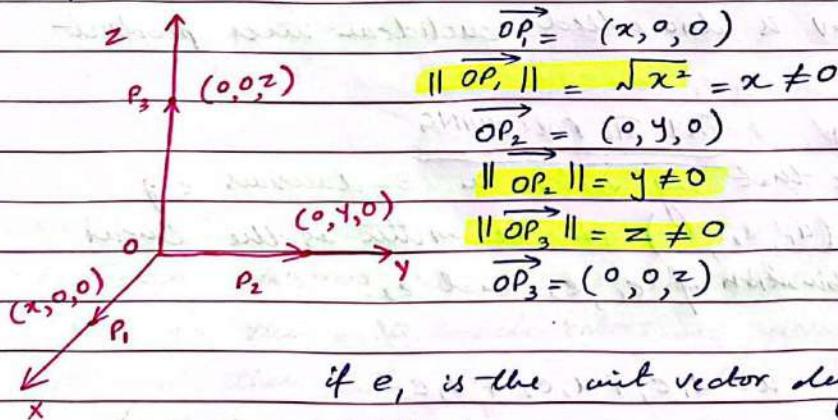


If $\|\mathbf{u}\|=1$ ie \mathbf{u} is a vector of norm 1 then \mathbf{u} is called a unit vector.

RESULT: In three dimensional space for $\mathbf{u} = (u_1, u_2, u_3)$

$$\|\mathbf{u}\| = \sqrt{u_1^2 + u_2^2 + u_3^2}$$

Unit Vectors along x, y and z axes in 3-dimensional space.



if e_i is the unit vector along x -axis then

$$e_1 = \frac{\overrightarrow{OP_1}}{\|\overrightarrow{OP_1}\|} = \frac{(x, 0, 0)}{\sqrt{x^2}} = \frac{(1, 0, 0)}{x}$$

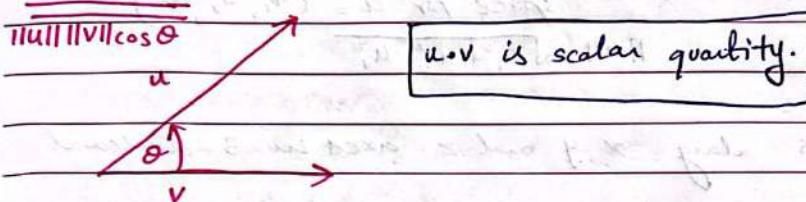
$\|e_1\|=1$, similarly unit vectors along y and z axes are given by $(0, 1, 0) = e_2$ and $(0, 0, 1) = e_3$

If u and v are vectors in 2 or 3-dimensional space and θ is angle between u and v then the dot product $u \cdot v$ is defined

by

$$u \cdot v = \begin{cases} \|u\| \|v\| \cos \theta & u \neq 0, v \neq 0 \\ 0 & u = 0, v = 0 \end{cases}$$

DOT PRODUCT



$u \cdot v$ is also called euclidean inner product

PROOF * TRY THE FOLLOWING

Prove that any vector in 3-dimensional e.g $u = (u_1, u_2, u_3)$ can be written as the linear combination of e_1, e_2 and e_3 .

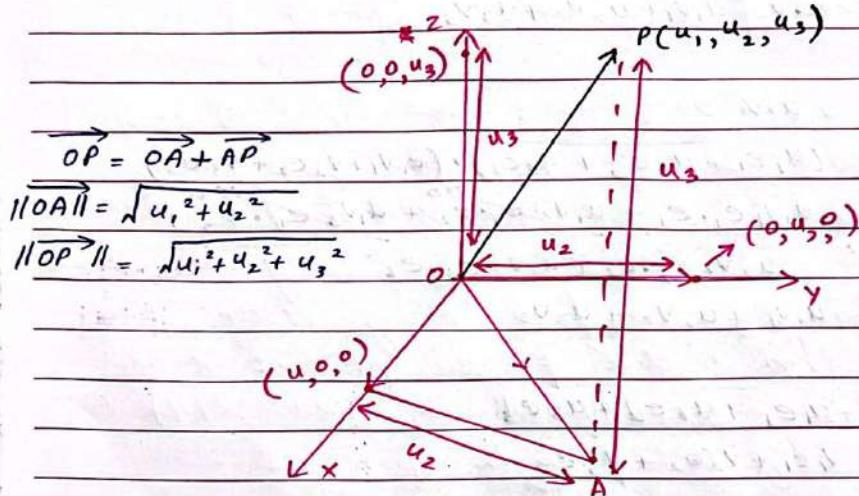
CHECK $u = u_1 e_1 + u_2 e_2 + u_3 e_3$

$$e_1 = (1, 0, 0), e_2 = (0, 1, 0), e_3 = (0, 0, 1).$$

DOT PRODUCT $\rightarrow \mathbf{a} \cdot \mathbf{b} = a_x b_x + a_y b_y + a_z b_z$

day / date:

$$\begin{aligned}\therefore u_1 e_1 + u_2 e_2 + u_3 e_3 \\&= u_1(1, 0, 0) + u_2(0, 1, 0) + u_3(0, 0, 1) \\&= (u_1, 0, 0) + (0, u_2, 0) + (0, 0, u_3) \\&= (u_1, u_2, u_3) = \overrightarrow{OP} = \mathbf{u}.\end{aligned}$$



ORTHOGONAL VECTORS

Dot product b/w two vectors is zero when the angle b/w them = 90° . Such vectors are perpendicular to each other and they are also called orthogonal vectors $\vec{u} \cdot \vec{v} = 0$

$\therefore e_1, e_2, e_3$ are orthogonal vectors

$$\boxed{e_i \cdot e_j = 1 \text{ if } i=j \quad i, j \in \{1, 2, 3\} \\ = 0 \text{ if } i \neq j}$$

$$\text{e.g. } e_1 \cdot e_2 = \|e_1\| \|e_2\| \cos 90^\circ = 0$$

$$e_1 \cdot e_1 = \|e_1\| \|e_1\| \cos 0^\circ = 1$$

PROVE THE FOLLOWING

$$\text{If } \mathbf{u} = (u_1, u_2, u_3) = u_1 \mathbf{e}_1 + u_2 \mathbf{e}_2 + u_3 \mathbf{e}_3$$

$$\mathbf{v} = (v_1, v_2, v_3) = v_1 \mathbf{e}_1 + v_2 \mathbf{e}_2 + v_3 \mathbf{e}_3$$

$$\text{then } \mathbf{u} \cdot \mathbf{v} = u_1 v_1 + u_2 v_2 + u_3 v_3$$

PROOF $\mathbf{u} \cdot \mathbf{v}$

$$= (u_1 \mathbf{e}_1 + u_2 \mathbf{e}_2 + u_3 \mathbf{e}_3) \cdot (v_1 \mathbf{e}_1 + v_2 \mathbf{e}_2 + v_3 \mathbf{e}_3)$$

$$= u_1 v_1 \mathbf{e}_1 \cdot \mathbf{e}_1 + u_1 v_2 \mathbf{e}_1 \cdot \mathbf{e}_2 + u_1 v_3 \mathbf{e}_1 \cdot \mathbf{e}_3 +$$

$$e_i \cdot e_j = 1 \quad \leftarrow u_2 v_2 \mathbf{e}_2 \cdot \mathbf{e}_2 + u_3 v_3 \mathbf{e}_3 \cdot \mathbf{e}_3 \quad \rightarrow e_i \cdot e_j = 0$$

$$\text{if } i=j = u_1 v_1 + u_2 v_2 + u_3 v_3$$

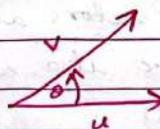
$$\text{if } i \neq j$$

$$\text{for } \mathbf{u} = u_1 \mathbf{e}_1 + u_2 \mathbf{e}_2 + u_3 \mathbf{e}_3$$

$$\mathbf{v} = v_1 \mathbf{e}_1 + v_2 \mathbf{e}_2 + v_3 \mathbf{e}_3$$

$$\mathbf{u} \cdot \mathbf{v} = u_1 v_1 + u_2 v_2 + u_3 v_3$$

$$= \|\mathbf{u}\| \|\mathbf{v}\| \cos \theta \quad \mathbf{u} \neq 0, \mathbf{v} \neq 0$$



$$\mathbf{u} \cdot \mathbf{u} = u_1^2 + u_2^2 + u_3^2 = \|\mathbf{u}\|^2$$

The dot product of vector with itself is the square of its norm.

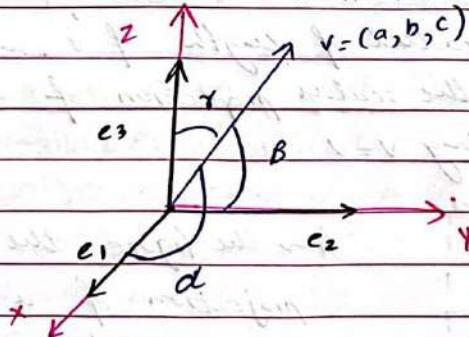
TRY THE FOLLOWING

① If $u = (2, -1, 1)$, $v = (1, 1, 2)$ then the angle b/w them $= 60^\circ$. $u \cdot v = \|u\| \|v\| \cos \theta$
 $2 + (-1) + 2 = 6 \cos \theta \Rightarrow \cos \theta = \frac{3}{6} \Rightarrow \theta = 60^\circ$

② Let e_1, e_2, e_3 be the unit vectors along x, y and z axes respectively in 3-dimensional space. If $v = (a, b, c)$ making angles α, β and γ with e_1, e_2 and e_3 respectively then

(i) find $\cos \alpha, \cos \beta, \cos \gamma$ prove that

$$(ii) \cos^2 \alpha + \cos^2 \beta + \cos^2 \gamma = 1$$



Note: α, β, γ are called the direction angles of v and $\cos \alpha, \cos \beta$ and $\cos \gamma$ are called the direction cosines of v .

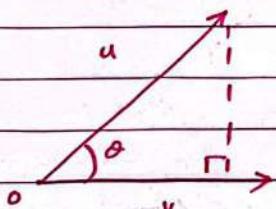
SCALAR & VECTOR PROJECTIONS.

Let $u = u_1 e_1 + u_2 e_2 + u_3 e_3$ we shall call $u_1 e_1$, $u_2 e_2$ and $u_3 e_3$ the vector components of u and the scalars u_1 , u_2 and u_3 the scalar components of x , y and z components of u .

In $u = u_1 e_1 + u_2 e_2 + u_3 e_3$ the scalar u_1 may be called the scalar projection of u on any vector whose direction is that of positive x -axis while the vector $u_1 e_1$ may be called vector projection of u on any vector whose direction is that of the positive x -axis therefore $u \cdot v = \|u\| \|v\| \cos \theta$ ($u \neq 0, v \neq 0$)

= product of length of v and the scalar projection of u on v or

of u along $v = v$

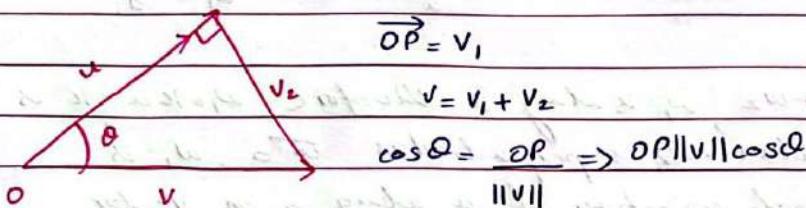


From the figure the scalar projection of u along v
 $= \|u\| \cos \theta = OP$

Similarly

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos\theta$$

= product of length of \mathbf{u} and the scalar projection of \mathbf{v} on \mathbf{u} .



which is the scalar projection of \mathbf{v} along \mathbf{u} .

Notice that $\mathbf{u} \cdot \mathbf{v} = \mathbf{v} \cdot \mathbf{u}$

SUMMARY

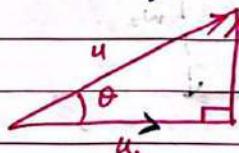
$$\mathbf{u} \cdot \mathbf{v} = u_1 v_1 + u_2 v_2 + u_3 v_3 = \|\mathbf{u}\| \|\mathbf{v}\| \cos\theta \quad \begin{matrix} u \neq 0 \\ v \neq 0 \end{matrix}$$

if vector components of a vector are at 90°

to each other they are also called

rectangular vector components

e.g. $\mathbf{u} = \mathbf{u}_1 + \mathbf{u}_2$, \mathbf{u}_1 and \mathbf{u}_2 are



rectangular vector components

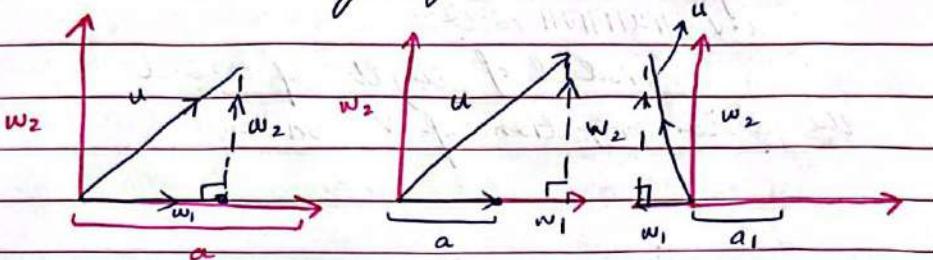
of \mathbf{u} and their norms are

$$\|\mathbf{u}_1\| = \|\mathbf{u}\| \cos\theta$$

$$\|\mathbf{u}_2\| = \|\mathbf{u}\| \sin\theta$$

$$0 < \theta < \pi/2, \mathbf{u}_1 = \mathbf{u}_2 \neq 0, \cos\theta > 0 \text{ and } \sin\theta > 0$$

consider the following figures:



$u = w_1 + w_2$, w_1 is along a therefore $w_1 = k a$ k is a scalar. w_2 is perpendicular to a . w_1 is orthogonal projection of u along a or vector component of u along a OR just $w_1 = \text{proj}_a u$

PROBLEM: How to find w_1 and w_2 ?

vector component of u along a .

If $a \neq 0$ then $w_1 = \text{proj}_a u = (u \cdot a) \cdot a$, $a \neq 0$
 $\|a\|^2$

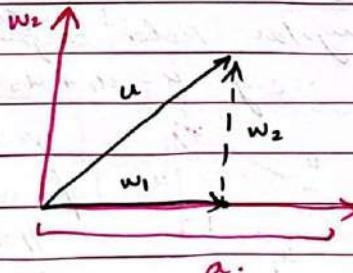
PROOF

$$u = w_1 + w_2$$

$$\therefore w_1 = k a$$

$$\therefore u = k a + w_2$$

$$\Rightarrow \frac{u}{2} = \text{proj}_a u - k a$$



$$\text{Proj}_a u \rightarrow \frac{(u \cdot a)}{\|a\|^2} \cdot a$$

perpendicular
 $w_1 \cdot a = 0$
day / date:

taking dot product with a ~~$w_1 \cdot a = 0$~~ $= (u - ka) \cdot a$

$$\Rightarrow u \cdot a - ka \cdot a = 0$$

$$\Rightarrow k = \frac{u \cdot a}{a \cdot a} = \frac{u \cdot a}{\|a\|^2} = k \quad a \cdot a = \|a\|^2$$

$$\therefore w_1 = ka = \frac{(u \cdot a)}{\|a\|^2} a$$

$$\therefore w_1 = \frac{(u \cdot a)}{\|a\|^2} a \quad \text{AND}$$

$$\text{Proj}_a u \leftarrow \frac{(u \cdot a)}{\|a\|^2} a$$

$$w_2 = u - \frac{(u \cdot a)}{\|a\|^2} a \rightarrow \text{vector component of } u \text{ orthogonal to } a.$$

$$w_2 = u - w_1$$

$$\text{also note that } w_2 \cdot a = 0$$

PROBLEM:

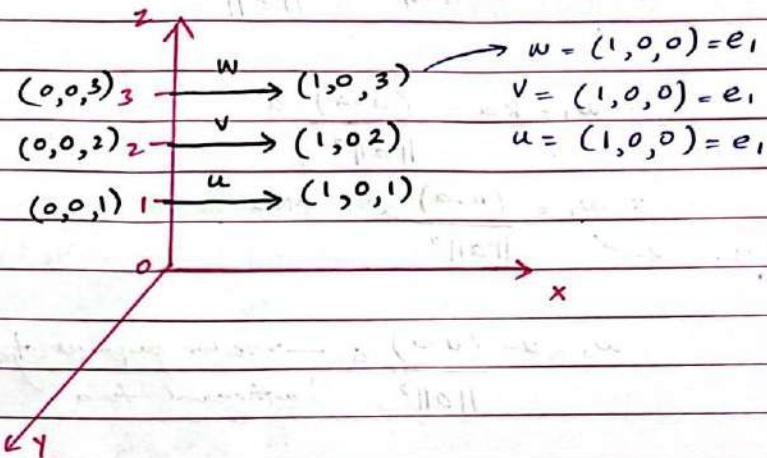
let $u = (2, -1, 3)$ $a = (4, -1, 2)$ find:

- ① vector component of u along a ($\text{Proj}_a u$) and
- ② the vector component of u orthogonal to a

Ans $(\text{Proj}_a u) = \left(\frac{20}{7}, \frac{-5}{7}, \frac{10}{7} \right)$ and the one orthogonal to $a = \left(-\frac{6}{7}, -\frac{2}{7}, \frac{11}{7} \right)$



Recall that two vectors are equal (equivalently) if they have same length and direction.
Consider the following vectors in xz -plane.



VECTOR SPACES

We shall state some properties, if satisfied by a set of objects will entitle those objects to be called vectors.

These new type of vectors will include various kinds of matrices and functions etc.

DEFINITION: let V be non-empty set of objects on which two operations are defined addition & multiplication by scalars (real numbers)

By addition we mean for $u, v \in V$ an object $u+v$ or $u \oplus v$

By scalar multiplication we mean for each scalar k and each object $u \in V$ an object ku , called the scalar multiple of u by k . or $k \cdot u$

$u+v \rightarrow$ vector addition

$ku \rightarrow$ scalar multiplication.

V is called a VECTOR SPACE for $u, v, w \in V$ and for all scalars k and L if the following 10 properties are satisfied:

① for all $u, v \in V$, $u+v \in V$ which means V is closed under VECTOR ADDITION



- (2) $u+v = v+u$ { COMMUTATIVE PROPERTY OF VECTOR ADDITION }
- (3) $(u) + (v+w) = (u+v) + w$ { ASSOCIATIVE PROPERTY OF VECTOR ADDITION }
- (4) There is an object 0 in V called ZERO VECTOR such that $0+u=u+0=u$ for all $u \in V$
 $0 \rightarrow$ additive identity or identity element with respect to vector addition.
- (5) for each $u \in V$ there is an object $-u \in V$ called the negative of u or additive inverse of u such that
 $u + (-u) = (-u) + u = 0$
- (6) if k is any scalar and u is any object in V then ku is in V which shows that V is closed under scalar multiplication.

day / date:

- ⑦ $k(u+v) = ku+kv$ [addition of vectors]
- ⑧ $(k+l)u = ku + lu$ [scalar multiplication]
- ⑨ $k(lu) = (kl)u$
- ⑩ $1u = u$

due to property no ④
ie vector space is a nonempty set.

$$d \quad d$$

Example: Prove that set V of all 2×2 matrices (vectors) with real entries is a vector space under the matrix addition (as vector addition) and matrix scalar multiplication.

NOTE: IN ORDER TO AVOID THE CONFUSION WITH ORDINARY VECTORS USE α, β, γ AS ELEMENTS OF V .

SOLUTION:

Let $\alpha = \begin{bmatrix} a_1 & a_2 \\ a_3 & a_4 \end{bmatrix}, \beta = \begin{bmatrix} b_1 & b_2 \\ b_3 & b_4 \end{bmatrix}$



$$\textcircled{1} \quad \alpha + \beta = \begin{bmatrix} a_1+b_1 & a_2+b_2 \\ a_3+b_3 & a_4+b_4 \end{bmatrix} \in V$$

$\therefore \alpha + \beta$ is also a 2×2 MATRIX

$$\textcircled{2} \quad \alpha + \beta = \begin{bmatrix} a_1+b_1 & a_2+b_2 \\ a_3+b_3 & a_4+b_4 \end{bmatrix}$$

$$= \begin{bmatrix} b_1+a_1 & b_2+a_2 \\ b_3+a_3 & b_4+a_4 \end{bmatrix} = \begin{bmatrix} b_1 & b_2 \\ b_3 & b_4 \end{bmatrix} + \begin{bmatrix} a_1 & a_2 \\ a_3 & a_4 \end{bmatrix}$$

$$= \beta + \alpha$$

$$\textcircled{3} \quad \alpha + (\beta + \gamma) = (\alpha + \beta) + \gamma \quad \text{obvious check this by taking } \gamma = \begin{bmatrix} c_1 & c_2 \\ c_3 & c_4 \end{bmatrix}$$

$$\textcircled{4} \quad 0 + \alpha = \alpha + 0 = \alpha \quad \text{obvious since}$$

$$0 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, \quad 0 + \alpha = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} a_1 & a_2 \\ a_3 & a_4 \end{bmatrix}$$

$$= \begin{bmatrix} a_1 & a_2 \\ a_3 & a_4 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \doteq \alpha + 0 = \begin{bmatrix} a_1 & a_2 \\ a_3 & a_4 \end{bmatrix}$$

$$(5) \alpha + (-\alpha) = (-\alpha) + \alpha = 0 \rightarrow (*)$$

for $-\alpha = \begin{bmatrix} -\alpha_1 & -\alpha_2 \\ -\alpha_3 & -\alpha_4 \end{bmatrix}$ (*) is satisfied

$\hookrightarrow 2 \times 2$ matrix

$$(6) k\alpha = k \begin{bmatrix} \alpha_1 & \alpha_2 \\ \alpha_3 & \alpha_4 \end{bmatrix} = \begin{bmatrix} k\alpha_1 & k\alpha_2 \\ k\alpha_3 & k\alpha_4 \end{bmatrix} \in V$$

$$(7) k(\alpha + \beta) = k\alpha + k\beta$$

$$(8) (k+L)\alpha = k\alpha + L\alpha$$

$$(9) k(L\alpha) = k \begin{bmatrix} L\alpha_1 & L\alpha_2 \\ L\alpha_3 & L\alpha_4 \end{bmatrix}$$

$$= k(L) \begin{bmatrix} \alpha_1 & \alpha_4 \\ \alpha_3 & \alpha_5 \end{bmatrix} = (kL)\alpha$$

$$(10) 1\alpha = \alpha.$$

TRY THE FOLLOWING

PROOF

Determine whether the set of all 2×2 matrices of the form $\begin{bmatrix} a & 1 \\ 1 & b \end{bmatrix}$ with matrix addition and scalar multiplication is vector space.



$\mathbb{R} \ni a$
 $(a, b) \in \mathbb{R}^2$

$(a_1, a_2, \dots, a_n) \in \mathbb{R}^n \Rightarrow n$ tuples

day / date:

$$A = \begin{bmatrix} a_{11} & 1 \\ 1 & b_{12} \end{bmatrix} \quad B = \begin{bmatrix} b_{11} & 1 \\ 1 & b_{22} \end{bmatrix}$$

$$A+B = \begin{bmatrix} a_{11}+b_{11} & 2 \\ 2 & b_{12}+b_{22} \end{bmatrix} \notin V$$

Not a vector space, axiom (1) fails

Example V = set of all real-valued functions

PROOF defined on the real time $(-\infty, \infty)$

Proving 10th property.

$$f \in V$$

$$f = x^2 + 3x + 1$$

$$1f = f \in V$$

Example $V = \mathbb{R}^n$

PROOF

$$\text{Assume } v_1 = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}, v_2 = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$v_1 + v_2 = \begin{bmatrix} a_1 + b_1 \\ a_2 + b_2 \\ \vdots \\ a_n + b_n \end{bmatrix} \in \mathbb{R}^n$$



$$(u+v) \cdot w = u \cdot w + v \cdot w$$

PROOF Let $u = (u_1, u_2, \dots, u_n)$, $v = (v_1, v_2, \dots, v_n)$

and $w = (w_1, w_2, \dots, w_n)$ then

$$\begin{aligned} (u+v) \cdot w &= (u_1 + v_1, u_2 + v_2, \dots, u_n + v_n) \cdot (w_1, w_2, \dots, w_n) \\ &= (u_1 + v_1)w_1 + (u_2 + v_2)w_2 + \dots + (u_n + v_n)w_n \\ &= (u_1w_1 + u_2w_2 + \dots + u_nw_n) + (v_1w_1 + v_2w_2 + \dots + v_nw_n) \\ &= u \cdot w + v \cdot w \end{aligned}$$

$v \cdot v \geq 0$. Further $v \cdot v = 0$ if and only if $v = 0$

PROOF

We have $v \cdot v = v_1^2 + v_2^2 + \dots + v_n^2 \geq 0$. Further equality

holds if and if $v_1 = v_2 = \dots = v_n = 0$ - that is, if and only if $v = 0$.

To solve the vector equation $x + u = v$ for x we can add $-u$ to both sides

$$(x+u) + (-u) = v + (-u)$$

$$x + (u - u) = v - u$$

$$x + 0 = v - u$$

$$x = v - u$$

PROPERTIES OF LENGTH IN R^n

(a) $\|u\| \geq 0$

(b) $\|u\| = 0$ if and only if $u = 0$

(c) $\|ku\| = |k| \|u\|$

(d) $\|u + v\| \leq \|u\| + \|v\|$

PROOF (c) If $u = (u_1, u_2, \dots, u_n)$ then $ku = (ku_1, ku_2, \dots, ku_n)$

so

$$\begin{aligned}\|ku\| &= \sqrt{(ku_1)^2 + (ku_2)^2 + \dots + (ku_n)^2} \\ &= |k| \sqrt{u_1^2 + u_2^2 + \dots + u_n^2}\end{aligned}$$

PROOF (d)

$$\|u + v\|^2 = (u + v) \cdot (u + v)$$

$$\|u + v\|^2 = (u \cdot u) + 2(u \cdot v) + v \cdot v$$

$$\|u + v\|^2 = \|u\|^2 + 2(u \cdot v) + \|v\|^2$$

$$\|u + v\|^2 = \|u\|^2 + 2\|u\|\|v\| + \|v\|^2$$

$$\|u + v\|^2 = (\|u\| + \|v\|)^2$$

$$\|u + v\| = \|u\| + \|v\|$$

day / date:

PROOF

$$\begin{aligned} d(\vec{u}, \vec{v}) &\leq d(\vec{u}, \vec{w}) + d(\vec{w}, \vec{v}) \\ d(\vec{u}, \vec{v}) &= d(\vec{u}, \vec{v}) \\ &= \|\vec{u} - \vec{v}\| \\ &= \|(\vec{u} - \vec{w}) + (\vec{w} - \vec{v})\| \\ &= \|\vec{u} - \vec{w}\| + \|\vec{w} - \vec{v}\| \\ d(\vec{u}, \vec{v}) &= d(\vec{u}, \vec{w}) + d(\vec{w}, \vec{v}) \end{aligned}$$

PROOF

$$\begin{aligned} \vec{u} \cdot \vec{v} &= \frac{1}{4} (\|\vec{u} + \vec{v}\|^2 - \|\vec{u} - \vec{v}\|^2) \\ \frac{1}{4} \|\vec{u} + \vec{v}\|^2 - \frac{1}{4} \|\vec{u} - \vec{v}\|^2 &= \frac{1}{4} (\vec{u} + \vec{v}) \cdot (\vec{u} + \vec{v}) - \frac{1}{4} (\vec{u} - \vec{v}) \cdot (\vec{u} - \vec{v}) \\ \frac{1}{4} ((\vec{u} \cdot \vec{u}) + 2\vec{u} \cdot \vec{v} + \vec{v} \cdot \vec{v}) - \frac{1}{4} ((\vec{u} \cdot \vec{u}) - 2\vec{u} \cdot \vec{v} + \vec{v} \cdot \vec{v}) \\ &= \frac{1}{4} (2\vec{u} \cdot \vec{v} + 2\vec{v} \cdot \vec{u}) \\ &= \frac{1}{4} (2(\vec{u} \cdot \vec{v} + \vec{v} \cdot \vec{u})) \\ &= \frac{1}{2} (\vec{u} \cdot \vec{v} + \vec{v} \cdot \vec{u}) \end{aligned}$$

$$\therefore \frac{1}{2} (\vec{u} \cdot \vec{v} + \vec{v} \cdot \vec{u}) \Rightarrow \vec{u} \cdot \vec{v}$$



PROOF :-

$$\begin{aligned}
 (\vec{A}\vec{u}) \cdot \vec{v} &= \vec{u} \cdot (\vec{A}^T \vec{v}) \\
 (\vec{A}\vec{u}) \cdot \vec{v} &= (\vec{A}\vec{u})^T \vec{v} \quad (\because \vec{u} \cdot \vec{v} = \vec{u}^T \vec{v}) \\
 &= \vec{u}^T \vec{A}^T \vec{v} \\
 &= \vec{u}^T (\vec{A}^T \vec{v}) \\
 (\vec{A}\vec{u}) \cdot \vec{v} &\stackrel{\text{Hence}}{=} \vec{u} \cdot (\vec{A}^T \vec{v})
 \end{aligned}$$

PROVE :-

$$\vec{0}\vec{u} = 0$$

PROOF $A \times 8 \Rightarrow (\kappa + \ell)\vec{u} = \kappa\vec{u} + \ell\vec{u}$

$$\vec{0}\vec{u} + \vec{0}\vec{u} = (0+0)\vec{u}$$

$$\vec{0}\vec{u} + \vec{0}\vec{u} = \vec{0}\vec{u}$$

$$A \times 4 \Rightarrow \vec{u} \in V \Leftrightarrow -\vec{u} \in V$$

$$\vec{0}\vec{u} + \vec{0}\vec{u} + (-\vec{0}\vec{u}) = \vec{0}\vec{u} + (-\vec{0}\vec{u})$$

$$A \times 3 \Rightarrow \vec{0}\vec{u} + (\vec{0}\vec{u} + (-\vec{0}\vec{u})) = \vec{0}\vec{u} + (-\vec{0}\vec{u})$$

$$A \times 5 \Rightarrow \vec{0}\vec{u} + 0 = 0$$

$$A \times 6 \Rightarrow \vec{0}\vec{u} = 0$$

PROVE :- $(-1)\vec{u} = -\vec{u} \quad \forall \vec{u} \in V$

PROOF $(-1)\vec{u} = -\vec{u}$

$$\Rightarrow \vec{u} + (-1)\vec{u} = 0$$

$$\vec{u} + (-1)\vec{u}$$

$$1\vec{u} + (-1)\vec{u}$$

$$(1 + (-1))\vec{u}$$

$$\vec{0}\vec{u} \Rightarrow 0$$

TRANSPOSE DOT PRODUCT

$$u = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix} \quad \text{and} \quad v = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$

$$v^T u = [v_1 \ v_2 \ \dots \ v_n] \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix} = [u_1 v_1 + u_2 v_2 + \dots + u_n v_n]$$

$$= [u \cdot v] = u \cdot v$$

$$u \cdot v = v^T u \quad \text{--- eq ⑦}$$

$$Au \cdot v = u \cdot A^T v \quad \text{--- eq ⑧}$$

PROOF

$$\begin{aligned} Au \cdot v &= \underbrace{(v^T A^T)}_{u \cdot v} u \quad \text{using eq ⑦} \\ u \cdot v &= (A^T v)^T u = u \cdot (A^T v) \\ &\quad v^T u \qquad u \cdot v \end{aligned}$$

$$u \cdot Av = A^T u \cdot v \quad \text{--- eq ⑨}$$

PROOF

$$u \cdot Av = (Av)^T u = v^T (A^T u) = A^T u \cdot v$$

day / date:

If u and v are vectors in R^n with the Euclidean product, then

$$u \cdot v = \frac{1}{4} \|u+v\|^2 - \frac{1}{4} \|u-v\|^2$$

(orthogonal vectors)

EUCLIDEAN DISTANCE

Euclidean Norm

$$\|u\| = (u \cdot u)^{1/2} = \sqrt{u_1^2 + u_2^2 + \dots + u_n^2}$$

Euclidean Distance

$$d(u, v) = \|u - v\| = \sqrt{(u_1 - v_1)^2 + (u_2 - v_2)^2 + \dots + (u_n - v_n)^2}$$

If u and v are orthogonal vectors in R^n with Euclidean inner product, then

$$\|u+v\|^2 = \|u\|^2 + \|v\|^2 \quad \text{Pythagorean theorem.}$$

CAUCHY-SCHWARZ

$$(u, v)^2 \leq (u, u)(v, v)$$

$$u \cdot v = (u^T v)^2 \leq u^T u v^T v = \|u\|^2 \|v\|^2$$



KAGHAZ
www.kaghaz.pk

Determine whether a given (v) set is a vector space under the given operations

The set of all pairs of real numbers (x, y) with operations

$$(x, y) + (x', y') = (x+x', y+y') \text{ AND}$$

VECTOR ADDITION

$$k(x, y) = (2kx, 2ky)$$

SCALAR MULTIPLICATION

Solution:

u v

$$\text{VECTOR } (1) \quad (x, y) + (x', y') = (x+x', y+y') \in v$$

\Rightarrow since this is also an ordered pair of REAL NUMBER

$u + v$

$$\begin{aligned} \checkmark (2) \quad (x, y) + (x', y') &= (x+x', y+y') \quad v+u \\ &= (x'+x, y'+y) = (x', y') + (x, y) \\ \Rightarrow u+v &= v+u \quad \text{for } u=(x, y) \end{aligned}$$

u $v = (x', y')$ $w = (x'', y'')$ AND

$$\begin{aligned} \checkmark (3) \quad (x, y) + [(x', y') + (x'', y'')] &\Rightarrow u + (v+w) \\ &= (x, y) + (x'+x'', y'+y'') \\ &= \{x + (x'+x''), y + (y'+y'')\} \\ &= \{(x+x') + x'', (y+y') + y''\} \\ &= (x+x', y+y') + (x'', y'') \\ (u+v)+w &= [(x, y) + (x', y')] + (x'', y'') \\ \Rightarrow u + (v+w) - (u+v) &+ w \\ \text{for } w &= (x'', y'') \dots \end{aligned}$$



$$u + 0 = 0 + u$$

day / date:

$$(4) \quad (x, y) + (0, 0) = (0, 0) + (x, y)$$
$$= (x+0, y+0) = (x, y)$$
$$\Rightarrow 0 = (0, 0)$$

✓ (5) If $u = (x, y)$, $-u = (x', y')$

then $u + (-u) = (-u) + u = 0$

$$(x, y) + (x', y') = (0, 0) \quad \text{--- (1)}$$

but $(x, y) + (x', y') = (\underline{x+x'}, \underline{y+y'})$
from (1) and (2) $= 0 \quad = 0$

$$x + x' = 0 \Rightarrow x' = -x$$

$$y + y' = 0 \Rightarrow y' = -y$$

$$\therefore -u = (-x, -y)$$

$$\therefore (x, y) + (-x, -y) = (-x, -y) + (x, y)$$
$$\Rightarrow (0, 0)$$

✓ (6) $k(x, y) = (2kx, 2ky) \in V$ SCALAR MULTIPLICATION

obvious $\Rightarrow ku \in V$

Ansar (7)

$$(7) \quad k(u+v) = k[(x, y) + (x', y')]$$
$$= k[(\underline{x+x'}, \underline{y+y'})]$$
$$= 2k(\underline{x+x'}) \quad 2k(\underline{y+y'})$$
$$= (2kx + 2kx', 2ky + 2ky')$$
$$= (2kx, 2ky) + (2kx', 2ky')$$
$$= k(x, y) + k(x', y') = ku + kv$$
$$k(u+v) = ku + kv$$



$$(8) (k+l)u = (k+l)(x, y)$$

$$= [2(k+l)x, 2(k+l)y] - \textcircled{1}$$

also $[2x(k+l), 2y(k+l)]$

$$ku + lu = k(x, y) + l(x, y)$$

$$= (2kx, 2ky) + (2lx, 2ly)$$

$$= [2(k+l)x, 2(k+l)y] - \textcircled{2}$$

from $\textcircled{1}$ and $\textcircled{2}$

$$(k+l)u = ku + lu$$

$$k[lu]$$

$$(9) k(lu) = k[l(x, y)]$$

$$= k(2lx, 2ly) = (4lkx, 4lky) - \textcircled{1}$$

$$(kl)u = (kl)(x, y) = (2klix, 2kly) - \textcircled{2}$$

from $\textcircled{1}$ and $\textcircled{2}$ $k(lu) \neq (kl)u$

so NOT a vector space.

$$(10) 1u = 1(x, y) = (2x, 2y) \neq (x, y)$$

$$1u \neq u$$

$$\therefore k(x, y) = (2kx, 2ky)$$

Axiom $\textcircled{9}$ and $\textcircled{10}$ fails.



SUBSPACES

A subset w of a vector space V is called a subspace of V if w is itself a vector space under the addition and scalar multiplication defined on V .

Theorem 5.2.1

If w is a set of one or more vectors from vector space V , then w is a subspace of the w is a subspace of V if and only if the following condition holds.

(a) If u and v are vectors in w then $\underline{u+v}$ is in w .

(b) If k is any scalar and u is any vector in w then $\underline{ku \in w}$

PROOF

If w is a subspace of V then all the vector space axioms or properties are satisfied including (1) and (6) which are same as (a) and (b) above.

Conversely, assume conditions (a) and (b) hold. Since they are vector space axioms 1 and 6 we need to show that other 8 are satisfied.



Axioms 2, 3, 7, 8, 9¹⁰ are satisfied including
 (ii) automatically satisfied by vectors in
 w since they are satisfied by all vectors in
 v .

Therefore to complete the proof, we need
 only to verify that axioms 4 and 5
 are satisfied by vectors in w .

Let u be any vector in w . By condition
 (b), $ku \in w$ for every scalar k .

Setting $k=0$, $ku=0u=0$
 but $ku \in w \Rightarrow 0 \in w$

and setting $k=-1$, it follows that
 $(-1)u = -u \in w$

RESULT w is subspace of v if and only
 if w is closed under addition and
 closed under multiplication.

day / date:

PROOF Prove that if $S = \{v_1, v_2, \dots, v_n\}$ is a basis for some vector space V , then any vector $u \in V$ can be written as linear combination.

Let u vector has two different representation

$$u = c_1 v_1 + \dots + c_n v_n \quad \text{--- (1)}$$

$$u = k_1 v_1 + \dots + k_n v_n \quad \text{--- (2)}$$

$$\text{eq (1)} - \text{eq (2)}$$

$$u - u = (c_1 v_1 + \dots + c_n v_n) - (k_1 v_1 + \dots + k_n v_n)$$

$$u = (c_1 - k_1) v_1 + \dots + (c_n - k_n) v_n$$

By using linear independent property

$$c_1 - k_1 = 0, c_2 - k_2 = 0, \dots, c_n - k_n = 0$$

$$c_i = k_i$$

Hence

u vector has exactly one representation.

PROOF Show that the set W of all polynomials of degree $\leq n$ (including zero polynomial) is a subspace of real-valued functions under addition and scalar multiplication.

$\rightarrow W$ is not an empty set since it

contains zero polynomial

$$0 = x^0 + 0x^1 + \dots + 0x^n$$

Let $u, v \in W$

so $u + v \in W$



day / date:

$$u = a_0 + a_1 x + \dots + a_n x^n \text{ and } v = b_0 + b_1 x + \dots + b_n x^n$$

so $\boxed{u+v}$ is

$$(a_0 + b_0) + (a_1 + b_1)x + \dots + (a_n + b_n)x^n$$

$$c_0 + c_1 x + \dots + c_n x^n \in W$$

Next

$k u \in W$, k is any scalar

$$k u = k(a_0 + a_1 x + \dots + a_n x^n)$$

$$= k a_0 + k a_1 x + \dots + k a_n x^n$$

$$= d_0 + d_1 x + \dots + d_n x^n \in W$$

hence it is a subspace.

PROOF Prove that for a homogeneous solution system of equations with an coefficient matrix, the solution space is subspace of R^n



KAGHAZ
www.kaghaz.pk

day / date:

Show that set W of all 2×2 matrices having zeros as the main diagonal is a subspace of vector space M_{22} !

Solution

$$\text{Let } u = \begin{bmatrix} 0 & a \\ b & 0 \end{bmatrix}, v = \begin{bmatrix} 0 & c \\ d & 0 \end{bmatrix}$$

$$u+v = \begin{bmatrix} 0 & a+c \\ b+d & 0 \end{bmatrix} \in W \text{ and } ku = \begin{bmatrix} 0 & ka \\ kb & 0 \end{bmatrix} \in W$$

since both $u+v$ and ku contain zeros as the main diagonal $\therefore W$ is a subspace of M_{22}

LINEAR COMBINATIONS

A vector w is called a LINEAR COMBINATION of the vectors v_1, v_2, \dots, v_n if it can be expressed in the form $w = k_1v_1 + k_2v_2 + \dots + k_nv_n$ where k_1, k_2, \dots, k_n are scalars.

EXAMPLES

① Any vector in 3-dimensional space can be expressed as linear combination of the vectors e_1, e_2 and e_3

$$u = u_1e_1 + u_2e_2 + u_3e_3$$

$$= u_1(1, 0, 0) + u_2(0, 1, 0) + u_3(0, 0, 1) \Rightarrow (u_1, u_2, u_3)$$



AT LEAST ONE
SOLUTION \Rightarrow LINEARLY
INDEPENDENT

day / date:

- ② Any polynomial of degree n can be written as a linear combination of the following $n+1$ elements $\{1, x, x^2, \dots, x^n\}$ as
- $$P(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_n x^n$$

- ③ Any two by two 2×2 matrix can be written as linear combination of

$$\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \text{ and } \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

→ If $S = \{v_1, v_2, \dots, v_n\}$ is a nonempty set of vectors then the vector equation

$$k_1 v_1 + k_2 v_2 + \dots + k_n v_n = 0$$
 has at least one solution

namely $k_1 = k_2 = k_n = 0$

If this is the only solution then S is called a linearly independent set and vectors in set S are called LINEARLY INDEPENDENT VECTORS.

Examples

- ① The set S given by $\{e_1, e_2, e_3\}$ is LINEARLY INDEPENDENT and the vectors e_1, e_2 and e_3 are linearly independent vectors.

CHECK :-

$$\text{Consider } k_1 e_1 + k_2 e_2 + k_3 e_3 = 0 = (0, 0, 0)$$

$$\Rightarrow k_1(1, 0, 0) + k_2(0, 1, 0) + k_3(0, 0, 1) = (0, 0, 0)$$

$$\Rightarrow (k_1, k_2, k_3) = (0, 0, 0)$$

$$\Rightarrow k_1 = 0, k_2 = 0, k_3 = 0$$

② $S = \{1, x, x^2, \dots, x^n\}$ is LINEARLY INDEPENDENT

since for

$$a_0 + a_1 x + a_2 x^2 + \dots + a_n x^n = 0 \quad \text{--- (1)}$$

$\forall x$

$$\Rightarrow a_0 = a_1 = \dots = a_n = 0$$

since ① is satisfied by infinite values of x
 otherwise ① has at most m distinct roots
 if all or ^{some} of the coefficients $\neq 0$

③ $S = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$ is LINEARLY INDEPENDENT

SOLUTION :-

$$k_1 \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + k_2 \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + k_3 \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} + k_4 \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = 0$$

$$\begin{bmatrix} k_1 & k_2 \\ k_3 & k_4 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\Rightarrow k_1 = k_2 = k_3 = k_4 = 0 \quad \therefore S \text{ is LINEARLY INDEPENDENT}$$

day / date:

If v_1, v_2, \dots, v_n are vectors in a vector space V and if every vector in V is expressible as a LINEAR COMBINATION of these vectors, then we say that v_1, v_2, \dots, v_n SPAN V .

Examples

→ 3-dimensional space

① $\{e_1, e_2, e_3\}$ spans \mathbb{R}^3 since $(u_1, u_2, u_3) = u = u_1 e_1 + u_2 e_2 + u_3 e_3 \quad \forall u \in \mathbb{R}^3$ for all.

② $\{1, x, x^2, \dots, x^n\}$ spans the vector space P_n since each polynomial $P(x)$ in P_n can be written as $P(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_n x^n$ (LINEAR COMBINATION)

③ $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$ span M_{22}

(all matrices of order 2) since

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} = a \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + b \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} + c \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} + d \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\forall \begin{bmatrix} a & b \\ c & d \end{bmatrix} \in M_{22}$$

day / date:

If V is any vector space and $S = \{v_1, v_2, \dots, v_n\}$ is a set of vectors in V , then S is called a basis for V if the following two conditions hold:

- (a) S is linearly independent
- (b) S spans V .

Let us consider some examples of sets which are bases i.e. they are LINEARLY INDEPENDENT as well as SPAN DIFFERENT VECTOR SPACES

Example (of BASES)

BASES \rightarrow Plural of Basis.

- ① $\{e_1, e_2, e_3\}$ is a basis for \mathbb{R}^3 because it's linearly independent as well as spans \mathbb{R}^3 similarly
- ② $\{1, x, x^2, \dots, x^n\}$ is basis for P_n and
- ③ $\left\{\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}\right\}$ is basis for M_{22}

DIMENSION: The dimensions of vector space V is defined to be the number of vectors in a basis for V .

REMARKS ① Dimension of $\mathbb{R}^3 = 3$ since there are three vectors in $\{e_1, e_2, e_3\}$

② Dimension of $P_n = n+1$

③ Dimension of $M_{22} = 4$

TRY THE FOLLOWING

check whether

$$\left\{ \begin{bmatrix} -1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$$

is a basis for M_{22} ?

HINT (i) First check that given matrices are linearly independent put

$$a_1 \begin{bmatrix} -1 & 1 \\ 0 & 0 \end{bmatrix} + a_2 \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} + a_3 \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} + a_4 \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \text{ and see if } a_1 = a_2 = a_3 = a_4 = 0$$

(ii) Take an arbitrary ~~any~~ element of M_{22}
as $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$ and check if it can be written
as a linear combination of given matrices.

For this put $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$

$$= k_1 \begin{bmatrix} -1 & 1 \\ 0 & 0 \end{bmatrix} + k_2 \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} + k_3 \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} + k_4 \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

and try to find k_1, k_2, k_3, k_4 in terms of
 a, b, c and d .

$$\left. \begin{array}{l} k_1 \\ k_2 \\ k_3 \\ k_4 \end{array} \right\} \text{unknowns} \quad \left. \begin{array}{l} a \\ b \\ c \\ d \end{array} \right\} \text{knowns.}$$

Answer: YES it is a basis

$$\therefore a_1 = a_2 = a_3 = a_4 = 0 \text{ and}$$

$$k_1 = \frac{b-a}{2}, \quad k_2 = \frac{a+b}{2}$$

$$k_3 = c$$

$$k_4 = d.$$

REMARK: A vector space may have more than one basis. But in all the bases (plural) the number of elements (vectors) are same. As we saw that

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

and $\left\{ \begin{bmatrix} -1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$

are bases for M_{22} and both contain 4 vectors
= dimension of M_{22}

Note: ① $\left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$

is called STANDARD BASES for M_{22} and

similarly $\{e_1, e_2, e_3\}$ and $\{1, x, x^2, \dots, x^n\}$
are standard bases for R^3 and P_n respectively.

② $\left\{ \begin{bmatrix} -1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$ is a basis

but not STANDARD BASIS for M_{22} .

REVISION

$$\textcircled{1} \quad \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$$

is called STANDARD BASES for $M_{2,2}$.

$$\textcircled{2} \quad \{e_1, e_2, e_3\} \text{ is STANDARD BASIS for } R^3$$

$$\textcircled{3} \quad \{1, x, x^2, \dots, x^n\} \text{ is STANDARD BASIS for } P_n$$

$$\textcircled{4} \quad \left\{ \begin{bmatrix} -1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\} \text{ is basis}$$

but not a standard basis for $M_{2,2}$ since

(first two elements are different from $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$) and

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

$$\textcircled{5} \quad \text{standard basis for } R^2 \text{ is } \{e_1, e_2\} \text{ where } e_1 = (1, 0), \\ e_2 = (0, 1)$$

RESULT: If $S = \{v_1, v_2, \dots, v_n\}$ is a set of n vectors
in an n -dimensional space V , then S is a basis
for V if either S spans V or S is linearly independent.



∴ An NPCLC is show that $\{(-3, 7), (5, 5)\}$ is basis for \mathbb{R}^2 .

solution two methods

① Since \mathbb{R}^2 is two dimensional space why?

Because $\{(1, 0), (0, 1)\}$ is the STANDARD BASIS

for \mathbb{R}^2 which contains two elements, therefore we only prove the given set to be linearly independent consider $k_1(-3, 7) + k_2(5, 5) = (0, 0)$

$$\Rightarrow -3k_1 + 5k_2 = 0$$

$$7k_1 + 5k_2 = 0$$

$$\Rightarrow \begin{bmatrix} -3 & 5 \\ 7 & 5 \end{bmatrix} \begin{bmatrix} k_1 \\ k_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \textcircled{1}$$

$$\det \begin{bmatrix} -3 & 5 \\ 7 & 5 \end{bmatrix} = -15 - 35 \neq 0$$

$$\Rightarrow \begin{bmatrix} -3 & 5 \\ 7 & 5 \end{bmatrix} \text{ is INVERTIBLE}$$

$$\textcircled{1} \Rightarrow \begin{bmatrix} k_1 \\ k_2 \end{bmatrix} = \begin{bmatrix} -3 & 5 \\ 7 & 5 \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow k_1 = k_2 = 0 \quad \therefore \text{Lin Indp and finally} \\ \{(-3, 7), (5, 5)\} \text{ is basis for } \mathbb{R}^2$$

(2) Since R^2 is two dimensional, we only prove that the given set spans R^2

consider

$$(x, y) = k_1(-3, 7) + k_2(5, 5)$$

$$\Rightarrow x = -3k_1 + 5k_2$$

$$y = 7k_1 + 5k_2$$

$$\Rightarrow \begin{bmatrix} -3 & 5 \\ 7 & 5 \end{bmatrix} \begin{bmatrix} k_1 \\ k_2 \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} k_1 \\ k_2 \end{bmatrix} = \begin{bmatrix} -3 & 5 \\ 7 & 5 \end{bmatrix}^{-1} \begin{bmatrix} x \\ y \end{bmatrix} \quad \text{--- (1)}$$

$\therefore \begin{bmatrix} -3 & 5 \\ 7 & 5 \end{bmatrix}$ is INVERTIBLE as

$$\det \begin{bmatrix} -3 & 5 \\ 7 & 5 \end{bmatrix} = -15 - 35 \neq 0$$

$$\therefore \begin{bmatrix} -3 & 5 \\ 7 & 5 \end{bmatrix}^{-1} = \frac{-1}{50} \begin{bmatrix} 5 & -5 \\ -7 & -3 \end{bmatrix}$$

$$(1) \Rightarrow \begin{bmatrix} k_1 \\ k_2 \end{bmatrix} = \frac{-1}{50} \begin{bmatrix} 5x - 5y \\ -7x - 3y \end{bmatrix}$$

$$k_1 = \frac{-1}{10} (x - y) \quad k_2 = \frac{1}{50} (-7x - 3y)$$

∴ Any element $(x, y) \in \mathbb{R}^2$ can be written as

LINEAR COMBINATION of $(-3, 7)$ and $(5, 5)$

∴ $\{(-3, 7), (5, 5)\}$ spans \mathbb{R}^2

NOTE $\{(-3, 7), (5, 5)\}$ is a basis but not

a standard basis since different from $\{(1, 0), (0, 1)\}$

Q11) Show that if $S : \{v_1, v_2, \dots, v_r\}$ is a linearly independent set of vectors, then so is every non-empty subset of S .

PROOF

Suppose that S has a linearly dependent subset T . Denote its vectors by w_1, \dots, w_m . Then

$$k_1w_1 + \dots + k_mw_m = 0$$

But if we let u_1, \dots, u_{n-m} denote the vectors which are in S but not in T , then

$$k_1w_1 + k_mw_m + 0u_1 + \dots + 0u_{n-m} = 0$$

Thus we have linear combination of vectors

v_1, \dots, v_n which equals 0. Since not all constants are zero, it follows that S is not linearly independent set of vectors, contrary to the hypothesis. That is, if S is linearly independent set, then so is every non-empty ^{sub}set, T .

Q13) Show that if $\{v_1, v_2, \dots, v_n\}$ is linearly dependent set of vectors in a vector space V and if v_{r+1}, \dots, v_n are any vectors in V , then $\{v_1, v_2, \dots, v_r, v_{r+1}, \dots, v_n\}$ is also linearly dependent.

B PROOF

Since $\{v_1, v_2, \dots, v_n\}$ is linearly dependent set of vectors, there exist constants c_1, c_2, \dots, c_r not all zero such that.

$$c_1 v_1 + c_2 v_2 + \dots + c_r v_r = 0$$

But then

$$c_1 v_1 + c_2 v_2 + \dots + c_r v_r + 0 v_{r+1} + \dots + 0 v_n = 0$$

The above equation implies that vectors v_1, \dots, v_n are linearly dependent.

Q15) Show that if $\{v_1, v_2\}$ is linearly independent and v_3 does not lie in span $\{v_1, v_2\}$ then $\{v_1, v_2, v_3\}$ is linearly independent.

PROOF:

Suppose that $\{v_1, v_2, v_3\}$ is linearly independent. Then there exists constants a, b and c not all zero such that

$$(*) \quad av_1 + bv_2 + cv_3 = 0$$

C.



CASE 1 $c=0$. Then (*) becomes

$$av_1 + bv_2 = 0$$

where not both a and b are zero.

CASE 2 $c \neq 0$. Then solving (*) for v_3 yields.

$$v_3 = -\frac{a}{c} v_1 - \frac{b}{c} v_2$$

This equation implies that v_3 is in $\text{span}\{v_1, v_2\}$ contrary to hypothesis. Thus $\{v_1, v_2, v_3\}$ is linearly independent.

RESULT: If $S = \{v_1, v_2, \dots, v_n\}$ is a basis for a vector space V then every vector v in V can be expressed in form

$$v = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$$

in exactly one way.

PROOF Let $v = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$ and $v = k_1 v_1 + k_2 v_2 + \dots + k_n v_n$ subtracting the second equation from the first gives

$$0 = (c_1 - k_1) v_1 + (c_2 - k_2) v_2 + \dots + (c_n - k_n) v_n$$

the linear independence of vectors in $\{v_1, v_2, \dots, v_n\}$ implies that

$$c_1 - k_1 = 0, c_2 - k_2 = 0, \dots, c_n - k_n = 0 \\ \Rightarrow c_1 = k_1, c_2 = k_2, \dots, c_n = k_n$$

- PREVIOUS RESULT
- ① $\{(1, 0), (0, 1)\}$ is standard basis for \mathbb{R}^2
 - ② $\{(-3, 7), (5, 5)\}$ is basis but not standard basis for \mathbb{R}^2

RECALL $(x, y) = x(1, 0) + y(0, 1)$ and

$$(x, y) = \frac{(y-x)}{10}(-3, 7) + \frac{(7x+3y)}{50}(5, 5)$$

HOMOGENOUS LINEAR SYSTEM

for $Ax=0$, exactly one of the following is true:

- ① system has only trivial solution ($x=0$ sol)
if A is invertible $\det(A) \neq 0$ (linear indp)
- ② system has infinitely many solutions in addition to trivial solution.

→ If $S = \{v_1, v_2, \dots, v_r\}$ is a non-empty set of vectors in vector equation.

$$k_1 v_1 + k_2 v_2 + \dots + k_r v_r = 0$$

if any one of scalars $k_i \neq 0$, $1 \leq i \leq r$ then S is linearly dependent and all vectors v_i are linearly dependent.

EXAMPLE check whether $\{(2, 2), (1, 1)\}$ is linearly dependent or independent?

$$k_1(2, 2) + k_2(1, 1) = (0, 0)$$

$$2k_1 + k_2 = 0$$

$$2k_1 + k_2 = 0$$

$$\begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} k_1 \\ k_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} : \det \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix} = 0$$

\therefore non-trivial solution exist hence given vectors are dependent.

NOTE Recall that if $Ax=0$ represents a homogeneous system of equation then infinitely many non-trivial solution exists if A is singular $\therefore A^{-1}$ does not exist

RECALL ① $\{(1, 1), (2, 2)\}$ is linearly dependent
 ② $\{(5, 5), (-3, 7)\}$ is " " independent.

GEOMETRIC INTERPRETATION. in \mathbb{R}^2

\rightarrow linearly independent if two vectors lie on SAME LINE.

\rightarrow linearly independent if they lie on different line.

NOTE: Also for any vector space V , the set $\{v_1, v_2\}$ is dependent if v_1, v_2 are scalar multiples of each other and $\{v_1, v_2\}$ is independent if and only if neither vector is scalar multiple.

~~lecture 16~~



RESULT A set S with two or more vectors is

- (a) linearly dependent if and only if at least one of the vectors in S is expressible as linear combination of other vectors in S .

e.g. in last ex $\{v_1, v_2, v_3\}$ is dependent
and $v_3 = v_2 + 3v_1$ i.e. v_3 is linear combination of v_1 and v_2 - scalar multiple.

- (b) " " is independent if and only if no vector in S is expressible as linear combination of other vectors in S .

e.g. $\{e_1, e_2, e_3\}$ is linearly independent
 \because none of e_1, e_2, e_3 is linear combination of other two.

$$\text{let } e_3 = k_1 e_1 + k_2 e_2$$

$$(0, 0, 1) = k_1(1, 0, 0) + k_2(0, 1, 0)$$

$$(0, 0, 1) = (k_1 k_2, 0) \text{ since NOT possible.}$$

LECTURE 17

RESULTS: ① If v_1 is a nonzero vector then

$\{v\}$ is always independent

$$\therefore kv_1 = 0 \Rightarrow k=0 \quad \because v_1 \neq 0$$

here $\{v\}$ is a set containing only one nonzero vector.

(2) If 0 is a zero vector then $\{0\}$ is always dependent
 $\because k0=0$ is satisfied by infinite nonzero values
of k .

→ The solution space of homogeneous system of equations $Ax=0$, which is a subspace of \mathbb{R}^n is called NULLSPACE of A

EXAMPLE: Nullspace of $A = \begin{bmatrix} 2 & 1 & 7 \\ -1 & 2 & -1 \\ 0 & 1 & 1 \end{bmatrix}$ is given by

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = t \begin{bmatrix} 3 \\ 1 \\ -1 \end{bmatrix}, t \text{ is any real number.}$$

→ The dimension of nullspace of A is called the NULLITY of A and is denoted by NULLITY(A).

Example: In above example $\text{NULLITY}(A)=1$ this is number of elements in the basis of nullspace of A i.e. number of elements in $\left\{ \begin{bmatrix} 3 \\ 1 \\ -1 \end{bmatrix} \right\}$ which is ONE.

DEFINITION for an $m \times n$ matrix

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}, \text{ the vectors (in } \mathbb{R}^n \text{)}$$

$$M_1 = (a_{11}, a_{12}, \dots, a_{1n})$$

$$M_2 = (a_{21}, a_{22}, \dots, a_{2n})$$

⋮

$$M_m = (a_{m1}, a_{m2}, \dots, a_{mn})$$

formed from rows of A are called **ROW VECTORS of A**
and vectors

$$C_1 = \begin{bmatrix} a_{11} \\ a_{12} \\ \vdots \\ a_{m1} \end{bmatrix}, C_2 = \begin{bmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix}, \dots, C_n = \begin{bmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{nn} \end{bmatrix} \text{ in } R^m$$

formed from columns of A are called **COLUMN VECTOR of A**

→ If A is an $m \times n$ matrix, then **subspace of R^m**
spanned by **row vectors of A** is called **row space**
of A and **subspace of R^n spanned by column vectors**
is called **column space of A** .

→ The dimensions of row space or column space of matrix
 A is called the **RANK of A** and denoted by **RANK(A)**

EXAMPLE

Consider the following $A = \begin{bmatrix} 2 & -1 & 0 & 3 \\ 1 & 2 & 5 & 0 \\ 7 & -1 & 5 & -8 \end{bmatrix}$

* largest possible value for $\text{rank}(A)$ is $\min(m, n)$
* min. no. of nullity of any matrix of order $m \times n$ is $\max(m, n) - \text{rank}(A)$

(*) find $\text{ROW}(\text{RANK})$ of A ?

$\text{ROW}(\text{RANK}) \rightarrow$ dimension of A row space or find
number of linearly independent row vectors so

$$\text{let } k_1 r_1 + k_2 r_2 + k_3 r_3 = 0 \rightarrow$$

$$k_1(2, -1, 0, 3) + k_2(1, 2, 5, -1) + k_3(7, -1, 5, 8) = (0, 0, 0, 0)$$

but last time it was proved that

$$k_1 = 3, k_2 = 1 \text{ and } k_3 = -1$$

\therefore 3 row vectors are linearly dependent

$$\therefore \text{ROW}(\text{RANK}) \neq 3$$

→ if scalar multiplied
other vector then linear dependent

now ignore $(7, -1, 5, 8)$ which is linear combination
of first two row vectors s.t. $3r_1 + r_2 = r_3$

$$a_1 r_1 + a_2 r_2 = 0$$

$$\Rightarrow a_1(2, -1, 0, 3) + a_2(1, 2, 5, -1) = (0, 0, 0, 0)$$

$$\Rightarrow 2a_1 + a_2 = 0$$

$$-a_1 + 2a_2 = 0$$

$$5a_2 = 0$$

$$3a_1 - a_2 = 0 \Rightarrow a_1 = a_2 = 0 \quad \text{first two are linearly}$$

independent.

$\text{ROW}(\text{RANK}) = 2$ of given matrix

$$\text{rank}(A^T) + \text{nullity}(A^T) = m \rightarrow \begin{matrix} \text{num of rows -} \\ \text{day/date:} \end{matrix}$$

NOTE ① Row space of A is spanned by $\{(2, -1, 0, 3), (1, 2, 5, -1)\}$ which is a linearly independent set and hence basis for row space

$\therefore \text{ROW(RANK)} = \text{RANK}(A) = 2 = \text{dimension of row space} = \text{no of linearly independent row vectors of } A$

② Similarly we can prove that column space of A is spanned by $\left\{ \begin{bmatrix} 2 \\ 1 \\ 7 \end{bmatrix}, \begin{bmatrix} -1 \\ 2 \\ -1 \end{bmatrix} \right\}$ which is linear independent set of vectors and hence basis of column space

$$\therefore \text{RANK}(A) = \text{COLUMN(RANK)} = 2$$

(b) find nullity (A)

for this we solve $AX=0$

$$\Rightarrow \begin{bmatrix} 2 & -1 & 0 & 3 \\ 1 & 2 & 5 & -1 \\ 7 & -1 & 5 & 8 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

3×4

augmented matrix is given by

$$\begin{bmatrix} 2 & -1 & 0 & 3 & 0 \\ 1 & 2 & 5 & -1 & 0 \\ 7 & -1 & 5 & 8 & 0 \end{bmatrix} \quad \begin{matrix} \text{now reduce this to} \\ \text{reduced echelon form} \end{matrix}$$



* system consistent if $\text{rank}(A) = \text{rank}(A|b) \geq 3$

day / date:

$$\sim \left[\begin{array}{ccccc} 1 & 2 & 5 & -1 & 0 \\ 2 & -1 & 0 & 3 & 0 \\ 7 & -1 & 5 & 8 & 0 \end{array} \right] \quad R_1 \leftrightarrow R_2$$

$$\sim \left[\begin{array}{ccccc} 1 & 2 & 5 & -1 & 0 \\ 0 & -5 & -10 & 5 & 0 \\ 0 & -15 & -30 & 15 & 0 \end{array} \right] \quad R_2 \rightarrow R_2 - 2R_1$$
$$R_3 \rightarrow R_3 - 7R_1$$

$$\sim \left[\begin{array}{ccccc} 1 & 2 & 5 & -1 & 0 \\ 0 & 1 & 2 & -1 & 0 \\ 0 & 1 & 2 & -1 & 0 \end{array} \right] \quad R_2 \rightarrow -1/5R_2$$
$$R_3 \rightarrow -1/15R_3$$

$$\sim \left[\begin{array}{ccccc} 1 & 2 & 5 & -1 & 0 \\ 0 & 1 & 2 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right] \quad R_3 \rightarrow R_3 - R_2$$

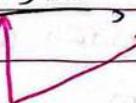
$$\sim \left[\begin{array}{cc|cc|c} 1 & 0 & 1 & 1 & 0 \\ 0 & 1 & 2 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right] \quad R_1 \rightarrow R_1 - 2R_2$$

$$\Rightarrow x_1 + (x_3 + x_4) = 0 \quad \dots \quad (2)$$

$$x_2 + 2x_3 - x_4 = 0 \quad \dots \quad (3)$$

→ 2 equations and four unknowns.

let $x_3 = t$, $x_4 = r$, $x_1 = -t - r$ from (2)


 $x_2 = -2x_3 + x_4$

$$x_4 = 2x_3 = s - 2t = x_2$$

FREE
VARIABLE

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -t-s \\ -2t+s \\ t \\ 3t+s \end{bmatrix} = t \begin{bmatrix} -1 \\ -2 \\ 1 \\ 0 \end{bmatrix} + s \begin{bmatrix} -1 \\ 1 \\ 0 \\ 3 \end{bmatrix}$$

\therefore basis for null space of $A = \left\{ \begin{bmatrix} -1 \\ -2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 0 \\ 3 \end{bmatrix} \right\}$

\therefore NULLITY of $A = 2$

= number of free variables = dimension of null space of A

$x_3 = t, x_4 = s \rightarrow$ free variables.

really
variables $\left\{ \begin{array}{l} x_1 = -t-s \\ x_2 = -2t+s \end{array} \right. \Rightarrow \text{RANK}(A) = 2$

DIMENSION THEOREM

$$2+2=4$$

If A has n columns then $\text{RANK}(A) + \text{NULLITY}(A) = n$

$n \rightarrow$ no of columns = total variables

Result

In $Ax=0$ $\left\{ \begin{array}{l} m \rightarrow \text{equations} \\ n \rightarrow \text{variables with } m \leq n \text{ and if there} \end{array} \right.$

$\left. \begin{array}{l} \text{are } r \text{ nonzero rows in the reduced row echelon} \\ \text{form of augmented matrix then number of free} \\ \text{variables are } n-r \\ 4-3=1 \end{array} \right.$

day / date:

Example

that for the following matrix

$$\begin{matrix} m=2 & n=4 \\ \left[\begin{array}{cccc} 2 & -1 & 0 & 3 \\ 1 & 2 & 5 & -1 \\ 7 & -1 & 5 & 8 \end{array} \right] & \end{matrix}$$

reduced row echelon form is

$$\left[\begin{array}{ccccc} 1 & 0 & 1 & 1 & 0 \\ 0 & 1 & 2 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right] \quad r=2$$

which contains two non-zero rows $\Rightarrow r=2$

$$\text{here } m=3 < 4 = n$$

$$\therefore n-r=4-2=2 \text{ free variables}$$

As we say that $x_3=t$ and $x_4=s$ were free variables and this $n=2$ also x_1 and x_2 were leading variables given by

$$x_1 = -t-s$$

$$x_2 = s-2t$$

and are $r=2 = \text{RANK}$ where r indicates non-zero rows in reduced row echelon of augmented matrix

ANOTHER METHOD TO FIND THE BASIS FOR ROW

SPACE OF MATRIX

In order to do this we refer to following

THEOREM The nonzero row vectors in any row echelon form of matrix form a basis for row space of that matrix.



EXAMPLE : The reduced row echelon form of

$$A = \begin{bmatrix} 2 & -1 & 0 & 3 \\ 1 & 2 & 5 & -1 \\ 7 & -1 & 5 & 8 \end{bmatrix} \text{ is given by}$$

$$R = \begin{bmatrix} 1 & 0 & 11 \\ 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad \text{the two nonzero rows}$$

vectors in R form basis for raw space

of A and R therefore the basis for raw space of R and A is given by

$$\{(1, 0, 11), (0, 1, 2, -1)\}$$

elementary row operations do not change the raw space of a matrix because

→ If A is an $m \times n$ matrix and B is raw equivalent to A then

- ① Raw space is a subspace of R^n
- ② If the raw operation is a raw interchange, the B and A have same raw vectors
- ③ If the raw operation is multiplication of raw by a nonzero scalar or addition of multiple of one raw to another then the raw vectors

, r'_m of B are linear combinations of thus they lie in the raw space of

(subspace of R^n) is closed under

addition and scalar multiplication.

METHOD TO FIND BASIS FOR COLUMN SPACE OF AN $m \times n$ MATRIX A.

If A is an $m \times n$ matrix and R is its row-reduced echelon form then the column vectors with leading 1's of row vectors form a basis for column space of R. and the corresponding column vectors in A form BASIS.

EXAMPLE:

for $A = \begin{bmatrix} 2 & -1 & 0 & 3 \\ 1 & 2 & 8 & -1 \\ 7 & -1 & 5 & 8 \end{bmatrix}$ the reduced row echelon

form is given by

$$R = \begin{bmatrix} 1 & 0 & 1 & 1 \\ 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad \text{RANK}(A) = 2$$

the first two column vectors in R which contain the leading 1's form the basis for the column space of R (but not A) and the corresponding column vectors in A form the basis for the column space of A (but not R) because elementary row operations usually change the column space.

REMARKS

① Basis for the column space for $R = \left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \right\}$

② Basis for the column space for $A = \left\{ \begin{pmatrix} 2 \\ 1 \\ 7 \end{pmatrix}, \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} \right\}$

③ Elementary Row operations can ~~not~~ change the column space:

let $A = \begin{bmatrix} 1 & 3 \\ 2 & 6 \end{bmatrix} \therefore$ the column space for A

consists of all scalar multiples of the first column vector $\begin{pmatrix} 1 \\ 2 \end{pmatrix}$

$$\text{Now } A \sim \begin{bmatrix} 1 & 3 \\ 0 & 0 \end{bmatrix}$$

\therefore The column space for $B = \begin{bmatrix} 1 & 3 \\ 0 & 0 \end{bmatrix}$ consists of all scalar multiples of first column vector ie $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$.
 This is not same as column space of A .

REMARK

If A is a matrix and R is its echelon form then the number of nonzero rows or number of column vectors that contain leading 1's in R is rank of matrix A .

LECTURE 19

day / date:

REVISION

If u and v are vectors in 2-space or 3-space and θ is the angle b/w u and v then the dot product or euclidean inner product $u \cdot v$ is defined by

$$u \cdot v = \begin{cases} \|u\| \|v\| \cos \theta & \\ 0 & \text{if } u=0 \text{ or } v=0 \end{cases}$$

$$(i) u \cdot v = v \cdot u$$

$$(ii) u \cdot (v+w) = u \cdot v + u \cdot w$$

$$(iii) k(u \cdot v) = (ku) \cdot v = u \cdot (kv)$$

$$(iv) v \cdot v > 0 \text{ if } v \neq 0 \text{ and } v \cdot v = 0 \text{ if } v = 0$$

$$v \cdot v = \|v^2\| > 0$$

The distance b/w two points is defined by

$$\|u-v\| = \|v-u\|$$

$$= [(v-u) \cdot (v-u)]^{1/2}$$

$$= \sqrt{(v_1-u_1)^2 + (v_2-u_2)^2 + (v_3-u_3)^2}$$

INNER PRODUCT

An inner product on a real vector space V is a real number $\langle u, v \rangle$ which satisfies following axioms for all $u, v, w \in V$ and all scalars k .



KAGHAZ
www.kaghaz.pk

- ① $\langle u, v \rangle = \langle v, u \rangle \rightarrow u \cdot v = v \cdot u$
- ② $\langle u+v, w \rangle = \langle w, u+v \rangle \rightarrow u \cdot (v+w) = u \cdot v + u \cdot w$
 $\langle u, w \rangle + \langle v, w \rangle = \langle w, u+v \rangle = \langle w, u \rangle + \langle w, v \rangle.$
- ③ $\langle ku, v \rangle = k \langle u, v \rangle \rightarrow k(u \cdot v) = (ku) \cdot v$
- ④ $\langle v, v \rangle \geq 0 \rightarrow v \cdot v > 0, v \neq 0$

→ A real vector space with an inner product is called a REAL INNER PRODUCT SPACE.

Example : If $u = (u_1, u_2, u_3)$ and $v = (v_1, v_2, v_3)$

are vectors in R^3 , then the formula

$$\langle u, v \rangle = u \cdot v = u_1 v_1 + u_2 v_2 + u_3 v_3$$

→ If V is an inner product space the norm of vector u in V is denoted by $\|u\|$ and is defined by

$$\|u\| = \sqrt{u \cdot u}$$

another notation of euclidean product :

for $u, v \in R^3$

$$u = \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}, v = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} \quad \langle u, v \rangle = u \cdot v = v^t u \text{ WHY?}$$

$$\therefore v^t u = [v_1 \ v_2 \ v_3] \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}$$



$$= u_1 v_1 + u_2 v_2 + u_3 v_3 = u \cdot v$$

$v^t \rightarrow$ TRANSPOSE OF V

TRY THE FOLLOWING

Let $f(x)$, $g = g(x)$ be two functions (continuous)

then decide whether

$$\langle f, g \rangle = \int_a^b f(x) g(x) dx \text{ defines an inner}$$

INNER
PRODUCT

product on $C[a, b]$

HINT check ① $\langle f, g \rangle = \langle g, f \rangle$

② $\langle f+g, s \rangle = \langle f, s \rangle + \langle g, s \rangle$

③ $\langle kf, g \rangle = k \langle f, g \rangle$

④ $\langle f, f \rangle \geq 0$ and $\langle f, f \rangle = 0$ if and

only if $f = 0$.

SOLUTION :- Let $f(x)$ and $g(x)$ be two functions.

such that $f, g \in C[a, b]$, { $f = f(x)$, $g = g(x)$ }

$C[a, b] \rightarrow$ all continuous functions defined
on interval $[a, b]$ consider

$$\textcircled{1} \quad \langle f, g \rangle = \int_a^b f(x) g(x) dx = \int_a^b g(x) f(x) dx = \langle g, f \rangle$$

$$\textcircled{2} \quad \langle f+g, s \rangle = \int_a^b [f(x) + g(x)] s(x) dx$$

$$= \int_a^b f(x) s(x) dx + \int_a^b g(x) s(x) dx$$

$$= \langle f, s \rangle + \langle g, s \rangle$$



day / date:

$$\langle g \rangle = \int_a^b k f(x) g(x) dx$$

$$= k \int_a^b f(x) g(x) dx = k \langle f, g \rangle$$

$$(4) \langle f, f \rangle = \int_a^b f(x) f(x) dx$$

$$= \int_a^b f^2(x) dx, \because f \in C[a, b]$$

$\therefore f^2$ is bounded by, $\because f^2(x) \geq 0$

$$\Rightarrow \min. f^2(x) = 0 \text{ AND}$$

$$\Rightarrow \max. f^2(x) = M$$

$$\Rightarrow 0 \leq f^2(x) \leq M$$

integrating

$$\int_a^b 0 dx \leq \int_a^b f^2(x) dx \leq \int_a^b M dx$$

$$\Rightarrow 0 \leq \int_a^b f^2(x) dx \leq M(b-a)$$

$$\int_a^b 0 dx = c \int_a^b = c - c = 0$$

NOTICE THAT

(1) $\int_a^b 0 dx$ is the area b/w x -axis
and x -axis and = 0

(2) $\int_a^b f^2(x) dx$ is the area b/w $y = f^2(x)$
and x -axis.

(3) $\int_a^b M dx = M(b-a)$ area b/w $f^2(x) = M$ and
 x -axis on $[a, b]$

$$\therefore \langle f, g \rangle = \int_a^b f(x) g(x) dx$$



KAGHAZ
www.kaghaz.pk

day / date:

DEFINITION : If $\langle \cdot, \cdot \rangle$ is an inner product of V and $u, v \in V$ then u and v are called

ORTHOGONAL VECTORS if

$$\langle u, v \rangle = 0, \text{ further if}$$

$\|u\| = \|v\| = 1$ then u and v are called **ORTHOGONAL VECTORS**.

TRY THE FOLLOWING

If $f(x) = \frac{1}{\sqrt{2}}$ and $g(x) = \sqrt{\frac{3}{2}} x$

(a) then according to the inner product defined in last example.

$$\langle f, g \rangle = \int_a^b f(x)g(x) dx$$

$f(x)$ and $g(x)$ are orthogonal and orthonormal on $[-1, 1]$??

IMPORTANT NOTES

① for $\langle f, g \rangle = \int_a^b f(x)g(x) dx$

$$\langle f, f \rangle^{1/2} = \|f\| = \sqrt{\int_0^b f^2(x) dx}$$



KAGHAZ
www.kaghazp.com

u and v are vectors from \mathbb{R}^2
and

$\langle u, v \rangle = u \cdot v = 0$ then u, v are
orthogonal as well as perpendicular.

- (3) If V is any inner product space and
 $\langle u, v \rangle \neq u \cdot v$ then $\langle u, v \rangle = 0$ means
 that u, v are orthogonal but not
 perpendicular.

~~LECTURE 20~~

PROBLEM: GRAM-SCHMIDT PROCESS

$V \rightarrow$ inner product space

Given: $\{u_1, u_2, \dots, u_n\}$ be any basis for V

then how to produce an orthogonal basis

$\{v_1, v_2, \dots, v_n\}$ for V ?

$$\text{i.e. } \langle v_i, v_j \rangle = 0, \quad i \neq j$$

$$1 \leq j \leq n, \quad 1 \leq i \leq n$$

which can be normalized to produce an
orthonormal basis i.e.

$$\left\{ \frac{v_1}{\|v_1\|}, \frac{v_2}{\|v_2\|}, \dots, \frac{v_n}{\|v_n\|} \right\}$$

i.e. norm of each vector = 1

Recall that for Euclidean inner product
(DOT PRODUCT)

$$\text{proj}_a u = \frac{(u \cdot a) a}{\|a\|^2} \text{ and vector projection}$$

(component) of u perpendicular to a is given by

$$u - \text{proj}_a u = u - \frac{(u \cdot a) a}{\|a\|^2}$$

similarly if v_1 and v_2 are orthogonal vectors in an inner product space V and $u \in V$ such that horizontal (\overrightarrow{OP}) projection of u lies along v_1 , then

$$v_2 = u - \frac{\langle u, v_1 \rangle v_1}{\|v_1\|^2}$$

PROOF

$$u = v_2 + \overrightarrow{OP}$$

$$\therefore u = k v_1 + v_2 = \overrightarrow{OP} + v_2$$

$$\therefore \overrightarrow{OP} = k v_1$$

$$\Rightarrow v_2 = u - \overrightarrow{OP} = u - k v_1 \quad \text{--- ①}$$

taking inner product on both sides

by v_1 , we get

$$\langle v_2, v_1 \rangle = \langle u, v_1 \rangle - \langle k v_1, v_1 \rangle$$

$$\Rightarrow \langle u, v_1 \rangle = \langle k v_1, v_1 \rangle$$

$$\langle u, v_1 \rangle = k \langle v_1, v_1 \rangle$$

$$k = \frac{\langle u, v_1 \rangle}{\|v_1\|^2}$$

day / date:

$$v_2 = u - \frac{\langle u, v_1 \rangle}{\|v_1\|^2} v_1$$

$\text{proj}_{v_1} u$

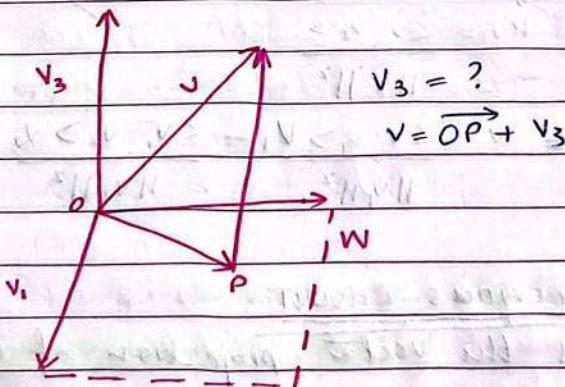
NOTE: Vectors $e_1 = (1, 0, 0)$ and $e_2 = (0, 1, 0)$

span the xy -plane because any vector in the xy -plane can be written as their linear combination.

$$(x, y, 0) = x(1, 0, 0) + y(0, 1, 0)$$

$$= xe_1 + ye_2$$

similarly if v_1 and v_2 are orthogonal vectors spanning a plane w as shown below,



v_3 is orthogonal to both v_1 and v_2 . \overrightarrow{OP} is the projection component of v in w . $\therefore \overrightarrow{OP}$ lies in w (spanned by v_1 and v_2)

\overrightarrow{OP} is linear combination of v_1 and v_2

$$\therefore \overrightarrow{OP} = k_1 v_1 + k_2 v_2 \text{ but } v = \overrightarrow{OP} + v_3 \Rightarrow v_3 = v - \overrightarrow{OP}$$

$$\Rightarrow v_3 = v - k_1 v_1 - k_2 v_2 \quad \text{--- (1)}$$

to find k_1 take inner product with v_1

$$\Rightarrow \underbrace{\langle v_3, v_1 \rangle}_{{v_i, v_j} \atop i \neq j = 0}^0 = \underbrace{\langle v, v_1 \rangle}_{\downarrow} - k_1 \underbrace{\langle v_1, v_1 \rangle}_{\|v_1\|^2} - k_2 \underbrace{\langle v_2, v_1 \rangle}_{\|v_2\|^2}$$

$$\Rightarrow k_1 = \frac{\langle v, v_1 \rangle}{\|v_1\|^2}$$

similarly to find k_2 take inner product with v_2

$$\Rightarrow \underbrace{\langle v_3, v_2 \rangle}_{{v_i, v_j} \atop i \neq j = 0}^0 = \underbrace{\langle v, v_2 \rangle}_{\downarrow} - k_1 \underbrace{\langle v_1, v_2 \rangle}_{\|v_1\|^2} - k_2 \underbrace{\langle v_2, v_2 \rangle}_{\|v_2\|^2}$$

$$\Rightarrow k_2 = \frac{\langle v, v_2 \rangle}{\|v_2\|^2}$$

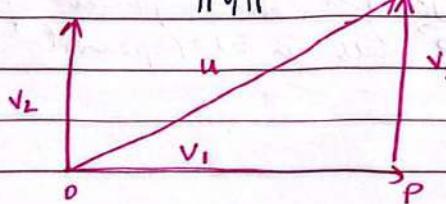
$$v_3 = v - \frac{\langle v, v_1 \rangle}{\|v_1\|^2} v_1 - \frac{\langle v, v_2 \rangle}{\|v_2\|^2} v_2$$

LECTURE 21

PREVIOUS RESULT

① If v_2 is the vector projection of u orthogonal to v_1 , then v_2 is given by

$$v_2 = u - \frac{\langle u, v_1 \rangle}{\|v_1\|^2} v_1 \text{ as shown}$$



$$u - \text{proj}_a u = u - \frac{\langle u, a \rangle}{\|a\|^2} a$$

$\|a\|^2 \text{ day / date:}$

② similarly, the vector projection v_3 of v which is orthogonal to both v_1 and v_2 is given by

$$v_3 = v - \frac{\langle v, v_1 \rangle}{\|v_1\|^2} v_1 - \frac{\langle v, v_2 \rangle}{\|v_2\|^2} v_2$$

→ (*)

SUMMARY

If V is an inner product space and $\{u_1, u_2, \dots, u_n\}$ is a basis for V then we can find the ~~operator~~ orthogonal basis by following these steps.

① let $u_1 = v_1 \rightarrow$ ①

② to find v_2 orthogonal to v_1

by computing the component of u_2 that is orthogonal to v_1

$$v_2 = u_2 - \text{proj}_{v_1} u_2 \quad \frac{\langle u_2, v_1 \rangle v_1}{\|v_1\|^2}$$

③ to find v_3 orthogonal to both v_1 and v_2 by computing component of u_3 orthogonal to plane spanned by v_1 and v_2 and is given by

$$v_3 = u_3 - \frac{\langle u_3, v_1 \rangle v_1}{\|v_1\|^2} - \frac{\langle u_3, v_2 \rangle v_2}{\|v_2\|^2}$$

replace v by u_3 in (*) in order
to get (3) so we obtained $v_1, v_2, v_3, \dots, v_n$
so on until we get v_n .

The preceding step by step construction for
converting an arbitrary basis into an orthogonal
basis is called. GRAM-SCHMIDT PROCESS.

Example :

Let vector space P_2 have inner product

$$\langle p, q \rangle = \int_{-1}^1 p(x)q(x) dx$$

apply the gram-schmidt process to
transform the standard basis $S = \{1, x, x^2\}$
into an arbitrary orthonormal basis.

SOLUTION:

$$\langle p, q \rangle = \int_{-1}^1 p(x)q(x) dx$$

$$S = \{1, x, x^2\}$$

$$u_1 = 1, u_2 = x, u_3 = x^2$$

$$① u_1 = v_1 = 1$$

$$② \|v_1\| = \|1\| = \sqrt{\langle v_1, v_1 \rangle} = \sqrt{\langle 1, 1 \rangle}$$

$$= \sqrt{\int_{-1}^1 1 dx} = \sqrt{\left[x \right]_{-1}^1} = \sqrt{2}$$



$$(3) v_2 = u_2 - \frac{\langle u_2, v_1 \rangle}{\|v_1\|^2} v_1$$

$$v_2 = x - \frac{\langle x, v_1 \rangle}{\|v_1\|^2} v_1 = x$$

$$(4) \|v_2\| = \|x\| = \left(\int_{-1}^1 x^2 dx \right)^{1/2}$$

$$\langle x, x \rangle^{1/2} = \sqrt{2/3}$$

$$(5) v_3 = u_3 - \frac{\langle u_3, v_1 \rangle}{\|v_1\|^2} v_1 - \frac{\langle u_3, v_2 \rangle}{\|v_2\|^2} v_2$$

$$\text{but } \langle u_3, v_1 \rangle = \int_{-1}^1 x^3 dx$$

$$= \frac{x^3}{3} \Big|_{-1}^1 = \frac{2}{3}$$

$$\text{and } \langle u_3, v_2 \rangle = \langle x^2, x \rangle$$

$$= \int_{-1}^1 x^3 dx = 0$$

$$\therefore v_3 = u_3 - \frac{\langle u_3, v_1 \rangle}{\|v_1\|^2} v_1 - \frac{\langle u_3, v_2 \rangle}{\|v_2\|^2} v_2$$

$$= x^2 - \frac{2}{3} \cdot \frac{1}{2} = x^2 - \frac{1}{3} = v_3$$

$$(6) \|v_3\| = \langle v_3, v_3 \rangle^{1/2}$$

$$\left(\int_{-1}^1 (x^2 - 1/3) dx \right) \Rightarrow \sqrt{8/45}$$

$$\textcircled{7} \quad \text{Required orthonormal basis is } = \left\{ \frac{\mathbf{v}_1}{\|\mathbf{v}_1\|}, \frac{\mathbf{v}_2}{\|\mathbf{v}_2\|}, \frac{\mathbf{v}_3}{\|\mathbf{v}_3\|} \right\}$$

$$= \left\{ \frac{1}{\sqrt{2}}, \sqrt{\frac{3}{2}}x, \left(x^2 - \frac{1}{3}\right) \sqrt{\frac{45}{8}} \right\}$$

$$= \left\{ \frac{1}{\sqrt{2}}, \sqrt{\frac{3}{2}}x, \frac{(3x^2 - 1)\sqrt{5}}{2\sqrt{2}} \right\}$$

TRY THE FOLLOWING

If $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is an orthogonal set of non-zero vectors in an inner product space, then S is linearly independent.

HINT Assume $k_1\mathbf{v}_1 + k_2\mathbf{v}_2 + \dots + k_n\mathbf{v}_n = 0$

and prove

$$k_1 = k_2 = \dots = k_n = 0$$

PROOF Let $k_1\mathbf{v}_1 + k_2\mathbf{v}_2 + \dots + k_n\mathbf{v}_n = 0$ taking inner product with \mathbf{v}_i on both sides,

$$\langle k_1\mathbf{v}_1 + k_2\mathbf{v}_2 + \dots + k_n\mathbf{v}_n, \mathbf{v}_i \rangle = 0$$

$$\begin{aligned} \therefore \langle 0, \mathbf{v}_i \rangle &= \langle 0 + 0, \mathbf{v}_i \rangle = \langle 0, \mathbf{v}_i \rangle + \langle 0, \mathbf{v}_i \rangle \\ &\Rightarrow \langle 0, \mathbf{v}_i \rangle = 0 \end{aligned}$$

$$\begin{aligned} k_1 \langle \mathbf{v}_1, \mathbf{v}_i \rangle + k_2 \langle \mathbf{v}_2, \mathbf{v}_i \rangle + \dots + k_i \langle \mathbf{v}_i, \mathbf{v}_i \rangle + \dots \\ + k_n \langle \mathbf{v}_n, \mathbf{v}_i \rangle = 0 \quad \text{--- (1)} \end{aligned}$$

day / date:

But ~~s~~ $S = \{v_1, v_2, \dots, v_n\}$ is an orthogonal set therefore $\langle v_i, v_j \rangle = 0$ when $i \neq j$
so that ① equation reduces to

$$k_i \langle v_i, v_i \rangle = 0 \text{ but } v_i \neq 0$$

$$\text{therefore } \langle v_i, v_i \rangle = \|v_i\|^2 > 0 \text{ so that}$$

$k_i = 0$. Since the subscript i is arbitrary we have $k_1 = k_2 = \dots = k_n = 0$; thus, S is linearly independent.

LECTURE 22

RESULT :

If A is an orthogonal matrix then

$$\det(A) = \pm 1$$

PROOF:

$$\because A^{-1} = A^T$$

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

$$\Rightarrow \frac{1}{\det(A)} = \det(A^T) = \det(A)$$

$$\det(A)$$

$$\Rightarrow \frac{1}{\det(A)} = \det(A)$$

$$\det(A)$$

$$\Rightarrow 1 = [\det(A)]^2$$

$$\Rightarrow \det(A) = \pm 1$$

NOTE : if $A^{-1} = A^T$
and $\det(A) = 1$ then

A is called proper
orthogonal matrix



EIGENVALUES / EIGENVECTORS.

NOTE: If A is an $n \times n$ is a column vector of $AX = b$ is also a column $n \times 1$. Both is true a relationship b/w X an

→ To FIND EIGENVALUES

$$\hookrightarrow A - \lambda I$$

$$\hookrightarrow \det(A - \lambda I) = 0$$

↪ find λ values.

→ TO FIND EIGENVECTOR

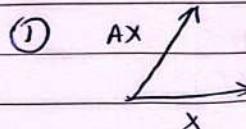
$$\hookrightarrow (A - \lambda I)X = 0$$

↪ insert λ values

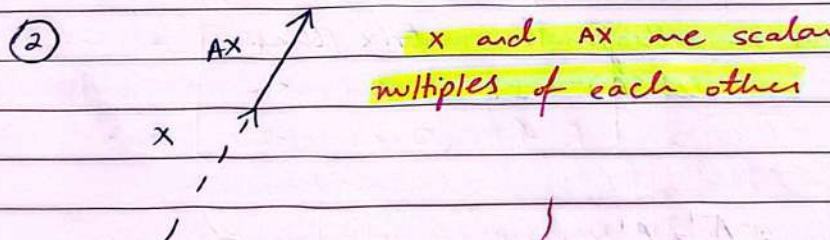
↪ form equations

↪ then find eigenvector.

SEE THE FOLLOWING FIGURE



no general "relationship between X and AX



X and AX are scalar

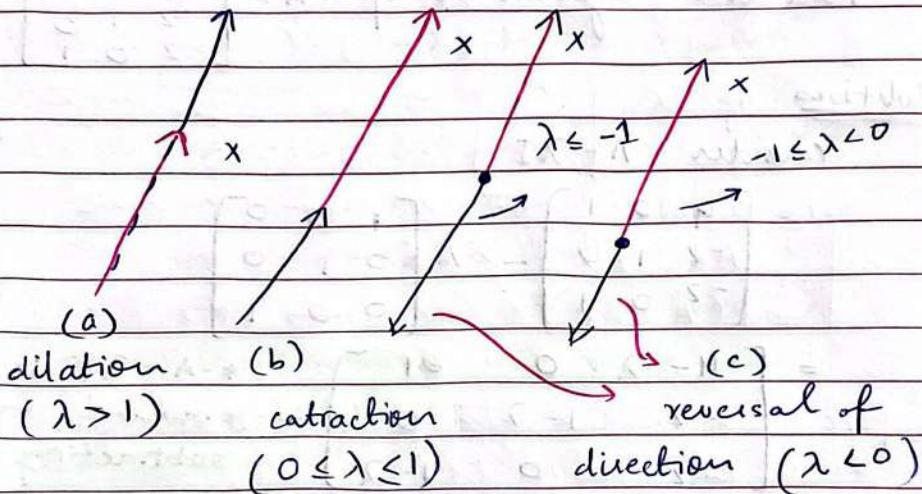
multiples of each other

DEFINITION If A is an $n \times n$ matrix then a non-zero vector X is called EIGENVECTOR of A if AX is scalar multiple of X . That is $AX = \lambda X$ for some scalar λ . The scalar λ is called EIGENVALUE of A and X is said to be EIGENVECTOR of A corresponding to λ .



GEOMETRIC INTERPRETATION

If $AX = \lambda X$, multiplication by A dilates X , contracts X , or reverses the direction of X , depending on the value of λ .

METHOD TO FIND EIGENVALUES

$$\therefore AX = \lambda X \Rightarrow AX = \lambda IX \quad \text{--- (1)}$$

where I is identity matrix

$$(1) \Rightarrow (A - \lambda I)X = 0 \quad \text{--- (2), } X \neq 0$$

since eigenvector is a non-zero vector therefore
for λ to be an eigenvalue of A, non-zero
(non-trivial) solutions of (2) exist if

$$\det(A - \lambda I) = 0 \rightarrow (*)$$

$(*)$ is called the characteristic equation of A

After solving (*), we get the eigenvalues (λ 's) of A .

EXAMPLE

Find the eigenvalues of $A = \begin{bmatrix} 4 & 0 & 1 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix}$

Solution

consider $A - \lambda I$

$$= \begin{bmatrix} 4 & 0 & 1 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 4-\lambda & 0 & 1 \\ -2 & 1-\lambda & 0 \\ -2 & 0 & 1-\lambda \end{bmatrix}$$

* $A - \lambda I$ is obtained by subtracting λ from **DIAGONAL ENTRIES** of A .

The characteristic equation of A is

$$\det(A - \lambda I) = 0, \quad A = \begin{bmatrix} 4 & 0 & 1 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix}$$

$$\Rightarrow \begin{vmatrix} 4-\lambda & 0 & 1 \\ -2 & 1-\lambda & 0 \\ -2 & 0 & 1-\lambda \end{vmatrix} = 0$$

$$\Rightarrow (4-\lambda)(1-\lambda)^2 + 2(1-\lambda) = 0$$

$$\quad \text{(+1)} (-2(1-\lambda))$$

$$\quad -\frac{2}{2}(1-\lambda)$$

$$(-2 \times 0) +$$

Order of Matrix = No of Eigen Values. day / date:

NOTE: This is an equation of degree 3 = order of A.

Therefore no. of eigenvalues of A are = order of A / 3

$$\textcircled{1} \Rightarrow (4-\lambda)(1-\lambda)^2 + 2(1-\lambda) = 0$$

$$\Rightarrow (1-\lambda)[(4-\lambda)(1-\lambda) + 2] = 0$$

$$1-\lambda = 0, \lambda = 1 \quad \textcircled{1}$$

$$(4-\lambda)(1-\lambda) + 2 = 0$$

$$\Rightarrow \lambda^2 - 5\lambda + 6 = 0$$

$$\Rightarrow (\lambda-2)(\lambda-3) = 0$$

$$\Rightarrow \lambda = 2, \lambda = 3 \quad \textcircled{2} \text{ and } \textcircled{3}$$

OR $\lambda_1 = 1, \lambda_2 = 2, \lambda_3 = 3$ are the eigenvalues of A.

NOTE: for an $n \times n$ matrix A, the characteristic equation $\det(A - \lambda I) = 0$ is of degree n and number of eigenvalues of A = n (a few may be repeated)

METHOD TO FIND EIGENVECTORS

EXAMPLE

Find the eigenvectors of $A = \begin{bmatrix} 4 & 0 & 1 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix}$

solution If $x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$ is an eigenvector of

A corresponding to λ then we have
 $AX = \lambda X = \lambda I X$
 $\Rightarrow (A - \lambda I)X = 0 \rightarrow (*)$

Therefore in order to obtain the eigenvectors we have to solve $(*)$

$$(*) \Rightarrow \begin{bmatrix} 4-\lambda & 0 & 1 \\ -2 & 1-\lambda & 0 \\ -2 & 0 & 1-\lambda \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Now we have to obtain the eigenvectors corresponding to $\lambda=1$, $\lambda=2$ and $\lambda=3$

STEP 1 eigenvector corresponding to $\lambda=1$
 putting $\lambda=1$ in ①

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

$$\Rightarrow 3x_1 + x_3 = 0 \quad -2x_1 = 0$$

$$\Rightarrow x_1 = x_3 = 0 \quad x_2 = t \quad \because t \neq 0$$

$$\therefore \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ t \\ 0 \end{bmatrix} = t \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

for $\lambda=1$, eigenvector is $\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ or any

of its nonzero multiple.

STEP 2 for $\lambda=2$

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

$$\Rightarrow 2x_1 + x_3 = 0 \Rightarrow x_3 = -2x_1$$

$$-2x_1 - x_2 = 0 \Rightarrow x_2 = -2x_1$$

$$-2x_1 - x_3 = 0 \quad x_1 = x_1$$



for $x_1 = t$, $x_2 = -2t$, $x_3 = -2t$

$$\Rightarrow \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} t \\ -2t \\ -2t \end{bmatrix} = t \begin{bmatrix} 1 \\ -2 \\ -2 \end{bmatrix}$$

\uparrow eigenvector.

STEP 3 finally for $\lambda = 3$, eigenvector is

$$\begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix} \text{ or any of multiples.}$$

~~LECTURE 23~~

EIGENSPACE: The eigenvectors corresponding to λ are the nonzero vectors in the solution space of $AX = \lambda X \Leftrightarrow (A - \lambda I)X = 0$. We call this solution space the eigenpace of A corresponding to λ .



QUESTION Find the bases for eigenspaces of

$$A = \begin{bmatrix} 4 & 0 & 1 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix}$$

Solution: Eigenvalues of A are 1, 2, 3
therefore there are THREE eigenspaces of

A corresponding to $\lambda = 1, 2$, and 3
so we proceed as follows-

① eigenspace corresponding to $\lambda = 1$ is given by

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = t \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \quad \text{eigenvector} = \text{eigenspace.}$$

→ BASIS.

② and ③ Therefore bases for eigenspaces corresponding to $\lambda = 2$ and $\lambda = 3$ are

given by $\left\{ \begin{bmatrix} 1 \\ -2 \\ -2 \end{bmatrix} \right\}$ and $\left\{ \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \right\}$

respectively.

NOTE: Dimension of each eigenspace of A described above = 1 \because each has only one basis vector.

LECTURE 24

* properties of diagonal and triangular matrices are diagonal entries.
day / date:

TRACE: If A is a square matrix then the trace of A is denoted by $\text{tr}(A)$ and is defined to be the sum of entries on the main diagonal of A .

EXAMPLE: for $A = \begin{bmatrix} 4 & 0 & 1 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix}$

$$\text{tr}(A) = 4 + 1 + 1 = 6$$

last time we saw that eigenvalues of $A = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix}$
consider

$$\lambda_1 + \lambda_2 + \lambda_3 = 1 + 2 + 3 = 6 = \text{tr}(A)$$

Result

If A is square matrix then $\text{tr}(A) = \text{sum of}$
it's eigenvalues also note that from ①

$$\det(A) = 4 + 1(2) = 6 = \lambda_1 \lambda_2 \lambda_3$$

$$\lambda_1 + \lambda_2 + \lambda_3 = \lambda_1 \lambda_2 \lambda_3$$

RESULT

If A is square matrix then $\det(A) = \text{product of it's}$
eigenvalues.



TRY THE FOLLOWING

(a) Show that the characteristic equation of 2×2 matrix A can be expressed as.

$$\lambda^2 - \text{tr}(A)\lambda + \det(A) = 0 \rightarrow ①$$

(b) If $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ then the solutions of characteristic equation of A are.

$$\lambda = \frac{1}{2} \left[(a+d) \pm \sqrt{(a-d)^2 + 4bc} \right]$$

NOTE: If λ_1, λ_2 are roots of ① then
 $\lambda_1 + \lambda_2 = \text{tr}(A)$, $\lambda_1 \lambda_2 = \det(A)$

RECALL

for α, β as roots of $ax^2 + bx + c = 0$, $a \neq 0$
 $\alpha + \beta = -\frac{b}{a}$ and $\alpha \beta = \frac{c}{a}$

DIAGONALIZATION

A square matrix A is called diagonalizable if there is an invertible matrix P such that $P^{-1}AP$ is diagonal matrix; the matrix P is said to diagonalize A.

$$AP = DP \quad * \text{if not to find day/date: } P^{-1}$$

EXAMPLE

$$= \begin{bmatrix} 2 & 0 & 1 \\ -1 & 0 & -1 \\ 0 & 1 & -1 \end{bmatrix} \xrightarrow{P^{-1}AP} \begin{bmatrix} 4 & 0 & 1 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix} \xrightarrow{\text{bases.}} \begin{bmatrix} 1 & 1 & 0 \\ -1 & -2 & 1 \\ -1 & -2 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 2 & 0 & 1 \\ -1 & 0 & -1 \\ 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} 3 & 2 & 0 \\ -3 & -4 & 1 \\ -3 & -4 & 0 \end{bmatrix} = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

P is the matrix having eigenvectors of A as its column vectors and $P^{-1}AP$ is a diagonal matrix having eigen values on main diagonal.

APPLICATION OF DIAGONALISATION.

The EigenVector Problem:

Given an $n \times n$ matrix A , does there exist a basis for \mathbb{R}^n consisting of eigenvectors of A ?

If A is an $n \times n$ matrix then the following are equivalent

(A) A is diagonalizable.

(B) A has n linearly independent EIGENVECTORS.



PROOF (b) \Rightarrow (a)

Assume that A has n linearly independent eigenvectors P_1, P_2, \dots, P_n , which with corresponding eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$

Consider the matrix P with P_1, P_2, \dots, P_n as its column vectors.

i.e

$$P = \begin{bmatrix} \downarrow P_1 & \downarrow P_2 & \downarrow P_n \\ P_{11} & P_{12} & \dots & P_{1n} \\ P_{21} & P_{22} & \dots & P_{2n} \\ \vdots & \vdots & & \vdots \\ P_{n1} & P_{n2} & \dots & P_{nn} \end{bmatrix}$$

now the columns of the product AP are

$$AP_1, AP_2, \dots, AP_n$$

EXTRA: Consider $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$ and $P = \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix}$,

$$AP = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} P_{11} & P_{12} \\ P_{21} & P_{22} \end{bmatrix}$$

$$\Rightarrow AP = \begin{bmatrix} a_{11}P_{11} + a_{12}P_{21} & a_{11}P_{12} + a_{12}P_{22} \\ a_{21}P_{11} + a_{22}P_{21} & a_{21}P_{12} + a_{22}P_{22} \end{bmatrix}$$

$\hookrightarrow P_{11}$ $\hookrightarrow P_{21}$

day / date:

First column of AP is $\begin{bmatrix} a_{11}P_{11} + a_{12}P_{21} \\ a_{21}P_{11} + a_{22}P_{21} \end{bmatrix}$

$$= \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} P_{11} \\ P_{21} \end{bmatrix} = AP_1$$

\times

\times

But $AP_1 = \lambda_1 P_1$, $AP_2 = \lambda_2 P_2$, ..., $AP_n = \lambda_n P_n$ so that

$$AP = \begin{bmatrix} \lambda_1 P_{11} & \lambda_2 P_{12} & \dots & \lambda_n P_{1n} \\ \lambda_1 P_{21} & \lambda_2 P_{22} & \dots & \lambda_n P_{2n} \\ \vdots & \vdots & & \vdots \\ \lambda_1 P_{n1} & \lambda_2 P_{n2} & \dots & \lambda_n P_{nn} \end{bmatrix}$$

$$= \begin{bmatrix} P_{11} & P_{12} & \dots & P_{1n} \\ P_{21} & P_{22} & \dots & P_{2n} \\ \vdots & \vdots & & \vdots \\ P_{n1} & P_{n2} & \dots & P_{nn} \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

$= PD$

$\Rightarrow AP = PD$ where D is the diagonal matrix having the eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ on the main diagonal. Since the column vectors of P are LINEARLY INDEPENDENT therefore

RANK (P) = n so that

$$\det(P) \neq 0$$



KAGHAZ
www.kaghaz.pk

day / date:

Recall that RANK is also defined as the highest order of nonzero determinant
∴ P is invertible

thus $AP=PD$ can be written as $P^{-1}AP=D$; that is A is diagonalizable.

Converse is also true.

$$\textcircled{A} \Rightarrow \textcircled{B}$$

If A is diagonalizable then A has n linearly independent eigenvectors. and they form a basis for \mathbb{R}^n .

- * unique values of eigenvalues in diagonal matrix then DIAGONALIZABLE.
- * if eigenvectors is not equal to order of matrix then it is NOT DIAGONALIZABLE.



KAGHAZ

www.kaghaz.p

ROTATION OF AXES.

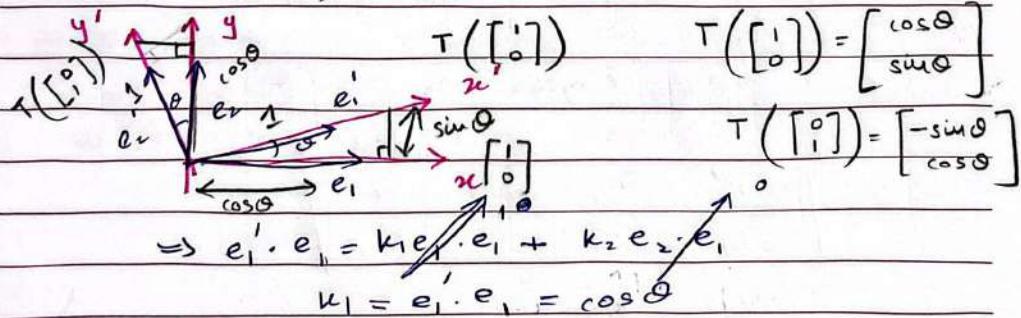
x -coordinate gives distance from y -axis, y -coord gives distance from x -axis. If the axes are rotated through an angle θ , then every point of plane has two representations:

(x, y) in original coordinate system and (x', y') in new system.

PROBLEM: What is the relationship b/w x and y of one coordinate system and x' and y' of other?

Consider the vector \overrightarrow{OP} which is given by $\overrightarrow{OP} = (x, y) = x e_1 + y e_2$ in original coordinate system and also

$\overrightarrow{OP} = (x', y') = x' e'_1 + y' e'_2$ in new coordinate system (e'_1 and e'_2 are also unit vectors)



similarly $e_2 = \sin\theta$

$$\therefore e'_1 = \cos\theta e_1 + \sin\theta e_2$$

similarly

$$e'_2 = -\sin\theta e_1 + \cos\theta e_2$$

$$\sim (-\sin\theta, \cos\theta)$$

$$\overrightarrow{OP} = x'e'_1 + y'e'_2$$

$$= x'(\cos\theta e_1 + \sin\theta e_2) + y'(-\sin\theta e_1 + \cos\theta e_2)$$

$$= (x'\cos\theta - y'\sin\theta)e_1 + (x'\sin\theta + y'\cos\theta)e_2$$

$$= xe_1 + ye_2 = \overrightarrow{OP} \quad \text{--- } ①.$$

shows that

$$x = x'\cos\theta - y'\sin\theta$$

$$y = x'\sin\theta + y'\cos\theta$$

$$\text{The matrix } \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} = R(\sin\theta) \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

which gives rotation through an angle θ
(counterclockwise) is called **ROTATION MATRIX**.

It's column vectors are new basic vector
ie $[e'_1, e'_2]$ and are orthonormal
with euclidean inner product.

day / date:

Also it's a vector and orthogonal with the Euclidean inner product.

NOTES:

① $RR^T = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$ into

$$\begin{bmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$
$$= \begin{bmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$$
$$= R^T R \Rightarrow R^T = R^{-1}$$

② $\det(R) = \cos^2\theta + \sin^2\theta = 1$

③ When there is no rotation then

$$R = I = \begin{bmatrix} [e_1] & [e_2] \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ for } \theta=0$$

$$\therefore \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x' \\ y \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} x' \\ y \end{bmatrix} = \begin{bmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$



KAGHAZ
www.kaghaz.pk

→ A square matrix A with the property
 $A^{-1} = A^T$ is said to be an orthogonal
orthogonal matrix.

OR

$$AA^T = A^T A = I$$



day / date:

LECTURE 25REVISION

Last time we saw that the eigenvalues

of $A = \begin{bmatrix} 4 & 0 & 1 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix}$ were distinct and = 1, 2, 3

corresponding eigenvectors are given by

$\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$, $\begin{bmatrix} 1 \\ -2 \\ -2 \end{bmatrix}$ and $\begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix}$ respectively which are

linearly independent \therefore they form a basis for \mathbb{R}^3 because A is diagonalizable as shown below:

$$\begin{aligned} P^{-1} A P &= \begin{bmatrix} 2 & 0 & 1 \\ 1 & 0 & -1 \\ 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} 4 & 0 & 1 \\ -2 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ -1 & -2 & 1 \\ -1 & -2 & 0 \end{bmatrix} \\ &= \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \end{aligned}$$

\rightarrow If v_1, v_2, \dots, v_n are eigenvectors of A corresponding to distinct eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ then $\{v_1, v_2, \dots, v_n\}$ is a linearly independent set.

+ distinct eigenvalues \rightarrow so eigenvectors are linearly independent.

KAGHAZ
www.kaghaz.pk

What will happen if eigenvalues are not distinct?

CONSIDER THE FOLLOWING EXAMPLE

Let $A = \begin{bmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{bmatrix}$ eigenvalues of A are

obtained by solving $\det(A - \lambda I) = 0$

$$= \begin{vmatrix} 4-\lambda & 2 & 2 \\ 2 & 4-\lambda & 2 \\ 2 & 2 & 4-\lambda \end{vmatrix} = 0$$

$$\Rightarrow (4-\lambda)[(4-\lambda)^2 - 4] - 2[2(4-\lambda) - 4] + 2[4 - 2(4-\lambda)] = 0$$

$$\Rightarrow (4-\lambda)[\lambda^2 - 8\lambda + 12] - 4[2(4-\lambda) - 4] = 0$$

$$\Rightarrow (4-\lambda)[(\lambda-6)(\lambda-2) - 8(4-\lambda-2)] = 0$$

$$\Rightarrow (4-\lambda)(\lambda-6)(\lambda-2) + 8(\lambda-2) = 0$$

$$\lambda = 2 \quad \text{OR} \quad (4-\lambda)(\lambda-6) + 8 = 0$$

$$-\lambda^2 + 10\lambda - 24 + 8 = 0$$

$$\lambda^2 - 10\lambda + 16 = 0$$

$$(\lambda-8)(\lambda-2) = 0$$

$$\lambda = 2, 2, 8$$

repeated twice.

EIGENVECTOR CORRESPONDING TO $\lambda = 8$

$$\begin{bmatrix} 4-8 & 2 & 2 \\ 2 & 4-8 & 2 \\ 2 & 2 & 4-8 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow -2x_1 + x_2 + x_3 = 0 \quad \text{--- (1)}$$

$$x_1 - 2x_2 + x_3 = 0 \quad \text{--- (2)}$$

$$x_1 + x_2 - 2x_3 = 0 \quad \text{--- (3)}$$

$$(1) + 2(2) \Rightarrow x_2 - 4x_2 + x_3 + 2x_3 = 0$$

$$\Rightarrow -3x_2 = -3x_3 \Rightarrow x_2 = x_3$$

$$(2) + 2(3) \Rightarrow 3x_1 + x_3 - 4x_3 = 0$$

$$\Rightarrow x_1 = x_3$$

$$\therefore \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_2 \\ x_2 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

CONSIDER FOR $\lambda = 2$

$$\begin{bmatrix} 4-2 & 2 & 2 \\ 2 & 4-2 & 2 \\ 2 & 2 & 4-2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow x_1 + x_2 + x_3 = 0$$

$$\Rightarrow x_1 = -t - s \quad \text{for } x_2 = t$$

$$x_3 = s$$

$$\therefore \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -t-s \\ t \\ s \end{bmatrix} = t \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} + s \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

∴ eigenvectors corresponding to $\lambda=2$ are of form given by (4) ie $\begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$ & $\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ form a

basis for the eigenspace corresponding to $\lambda=2$ therefore they are linearly independent.

Therefore eigenvectors corresponding to $\lambda=2$ are $\begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$ (for $s=0, t=1$) & $\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ (for $s=1, t=0$)

or linear combination of $\begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$. Notice

that $\lambda=2$ is repeated twice and the corresponding eigenspace is also two dimensional.

$\therefore \text{BASIS} = \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$ in this case

A is diagonalizable and we could easily check that $P^{-1}AP = D$

ie $P^{-1}AP$

$$= \frac{1}{3} \begin{bmatrix} 1 & 1 & 1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{bmatrix} \begin{bmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{bmatrix} \begin{bmatrix} 1 & -1 & -1 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 8 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$

NOTE:- If an eigenvalue λ of A is repeated k times and the eigenspace corresponding to λ is k -dimensional then the set consisting of basis vectors $\{v_1, \dots, v_k\}$ is linearly independent and they are also eigenvectors corresponding to λ as we saw in last example.

EXAMPLE:- Eigenvalues of A are

$$\begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 1 & 2 \end{bmatrix} = A \text{ are } \lambda_1 = 3, \lambda_2 = \lambda_3 = 2$$

↳ triangular matrix. It's eigenvalues are main diagonal entries



for $\lambda = 2$:- eigenvectors

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow x_1 = x_2 = 0 \\ x_3 = t \text{ (SAY)}$$

The eigenspace corresponding to $\lambda = 2$ is one dimensional but $\lambda = 2$ is repeated twice, so A is not diagonalizable as only two out of three eigenvectors of A are linearly independent.

$$\therefore \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = t \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \quad 1 \neq 2$$

LECTURE 26

ORTHOGONAL DIAGONALIZATION.

The ORTHONORMAL eigenvector problem:

Given an $n \times n$ matrix A , does there exist an orthonormal basis for \mathbb{R}^n with the Euclidean inner product consisting of eigenvectors of A ?

DEFINITION: A square matrix A is called **ORTHOGONALLY DIAGONALIZABLE** if there is an **ORTHOGONAL MATRIX P** such that

$P^{-1}AP = P^t AP$ is a diagonal matrix,
the matrix P is said to **ORTHOGONALLY DIAGONALIZE** A .

NOTE: Recall that for an **ORTHOGONAL MATRIX P** we have

$$P^t = P^{-1} \quad \text{OR} \quad P^t P = P P^t = I$$

TRY THE FOLLOWING

If A is orthogonally diagonalizable then prove
that A is a symmetric matrix.

* \star

PROOF $\therefore P^t AP = D$ where D is diagonal matrix
and $P^t P = I$

$$\therefore \underbrace{P P^t}_{I} A \underbrace{P P^t}_{I} = P D P^t$$

$$\Rightarrow A = P D P^t \quad \text{--- (1)}$$

$$\Rightarrow A^t = (P D P^t)^t = (P^t)^t \boxed{D^t} P^t$$

$$\Rightarrow P D P^t = A \text{ from (1)}$$

$$\Rightarrow A^t = A$$

NOTE: SYMMETRIC is always diagonalizable.

PROBLEM: find an orthogonal matrix P that diagonalizes

$$A = \begin{bmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{bmatrix}, A^T = A$$

STEPS ① find the eigenvalues of A they are given by $\lambda_1 = \lambda_2 = 2$ and $\lambda_3 = 8$

② find the basis for the eigenspace corresponding to $\lambda = 2$ and is given by

$$\{u_1, u_2\} = \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$$

NOTE: $u_1 \cdot u_2 = 1 \neq 0$ so u_1 is not orthogonal to u_2 . *if they are orthogonal then only make them unit vector.

③ Apply Gram-Schmidt process to $\{u_1, u_2\}$ to get an orthonormal basis i.e. $\left\{ \frac{v_1}{\|v_1\|}, \frac{v_2}{\|v_2\|} \right\}$

$$v_1 = u_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \frac{v_1}{\|v_1\|} = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} = w_1 (\text{sny})$$

$$v_2 = u_2 - \frac{(u_2 \cdot v_1) v_1}{\|v_1\|^2} = \frac{1}{2} \begin{bmatrix} -1 \\ -1 \\ 2 \end{bmatrix} \quad (\text{CHECK})$$

$$\|v_2\| = \sqrt{\frac{6}{2}} \rightarrow \frac{v_2}{\|v_2\|} = \frac{1}{\sqrt{6}} (-1, -1, 2) = w_2$$

④ find the basis for eigenspace corresponding to $\lambda = 8$. In this case

$$\text{BASIS} = \{u_3\} = \left\{ \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\}$$

⑤ Apply gram-schmidt process to u_3 to get

$$w_3 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \quad v_3 = u_3$$

NOTE: No need to find v_3 by using u_1 and u_2 in step ③ "eigenspaces are different".

⑥ finally using w_1, w_2 and w_3 as column vectors we obtain

$$P = \begin{bmatrix} -1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ +1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ 0 & 2/\sqrt{6} & 1/\sqrt{3} \end{bmatrix}$$

for $\lambda = 2$

which orthogonally diagonalizes A.



CHECK

PPT

$$= \begin{bmatrix} -1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ 1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ 0 & 2/\sqrt{6} & 1/\sqrt{3} \end{bmatrix} \begin{bmatrix} -1/\sqrt{2} & 1/\sqrt{2} & 0 \\ -1/\sqrt{6} & -1/\sqrt{6} & 2/\sqrt{6} \\ 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{3} \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \therefore P \text{ is an orthogonal matrix}$$

further P^TAP

$$= \begin{bmatrix} -1/\sqrt{2} & 1/\sqrt{2} & 0 \\ -1/\sqrt{6} & -1/\sqrt{6} & 2/\sqrt{6} \\ 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{3} \end{bmatrix} \begin{bmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{bmatrix} \begin{bmatrix} -1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ 1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ 0 & 2/\sqrt{6} & 1/\sqrt{3} \end{bmatrix}$$

$$= \begin{bmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 8 \end{bmatrix} = D$$

for the symmetric matrix $A = \begin{bmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{bmatrix}$, basis

for the eigenspace which corresponds to $\lambda = 2$
is given by $\{u_1, u_2\} = \{(-1, 1, 0), (-1, 0, 1)\}$ and

the basis for the eigenspace corresponding to $\lambda = 8 = \{u_3\} = \{(1, 1, 1)\}$ notice that

$$u_1 \cdot u_3 = (-1, 1, 0) \cdot (1, 1, 1) = 0.$$

$$\text{and } u_2 \cdot u_3 = (-1, 0, 1) \cdot (1, 1, 1) = 0$$

→ If A is symmetric matrix then eigenvectors from different eigenspaces are orthogonal.

→ If A and B are square matrix we say B is similar to A if there is an invertible matrix P such that $B = P^{-1}AP$

LINEAR TRANSFORMATIONS.

$$y = f(x)$$

dependent variable independent variable
 ✓
 BOTH ARE SCALAR

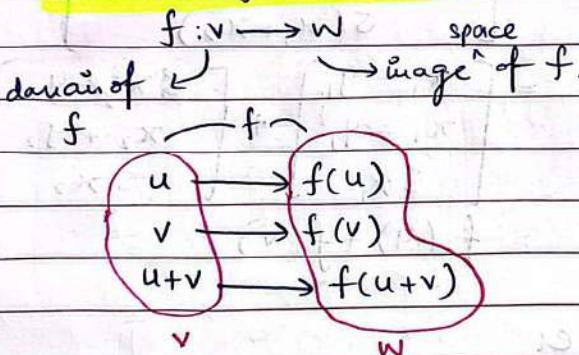
We shall begin the study of functions of the form $w = f(v)$ where the independent variable v and dependent variable w are both vectors.

We shall study functions which are called **LINEAR TRANSFORMATIONS**. Consider the following

→ If $f: V \rightarrow W$ is a function from the vector space V into vector space W , then f is called a **LINEAR TRANSFORMATION** if

(a) $f(u+v) = f(u) + f(v)$ for all $u, v \in V$

(b) $f(ku) = kf(u)$ for all $u \in V$ and all scalars k



EXAMPLE

Let $f: \mathbb{R}^2 \rightarrow \mathbb{R}^3$ be given by

$$f\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} x-y \\ x+y \\ 5x \end{bmatrix} \rightarrow (\star)$$

is f linear? or is f a linear transformation / mapping from $\mathbb{R}^2 \rightarrow \mathbb{R}^3$?

$$(0) \times \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

SOLUTION

let $u = \begin{bmatrix} x_1 \\ y_1 \end{bmatrix}, v = \begin{bmatrix} x_2 \\ y_2 \end{bmatrix}$

$$f(u+v) = f\left(\begin{bmatrix} x_1 \\ y_1 \end{bmatrix} + \begin{bmatrix} x_2 \\ y_2 \end{bmatrix}\right)$$

$$= f\left(\begin{bmatrix} x_1 + x_2 \\ y_1 + y_2 \end{bmatrix}\right)$$

$$= \begin{bmatrix} (x_1 + x_2) - (y_1 + y_2) \\ (x_1 + x_2) + (y_1 + y_2) \\ 5(x_1 + x_2) \end{bmatrix}$$

$$= \begin{bmatrix} x_1 - y_1 \\ x_1 + y_1 \\ 5x_1 \end{bmatrix} + \begin{bmatrix} 5x_2 - y_2 \\ x_2 + y_2 \\ 5x_2 \end{bmatrix}$$

$$= f(u) + f(v)$$

Now consider

$$f(ku) = f\left(k\begin{bmatrix} x \\ y \end{bmatrix}\right)$$

$$= f\left(\begin{bmatrix} kx \\ ky \end{bmatrix}\right)$$

$$= \begin{bmatrix} kx - ky \\ kx + ky \\ 5x \end{bmatrix} \quad \therefore f\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} x - y \\ x + y \\ 5x \end{bmatrix}$$

$$= k\begin{bmatrix} x - y \\ x + y \\ 5x \end{bmatrix} = kf(u)$$

from ① and ② f is a linear transformation
 from $\mathbb{R}^2 \rightarrow \mathbb{R}^3$

TRY THE FOLLOWING

Let $D: W \rightarrow V$ be the transformation that
 maps $f = f(x)$ into its derivative that is

$$D(f) = f'(x)$$

is D linear?

SOLUTION

$$\begin{aligned} D(f+g) &= (f(x)+g(x))' \\ &= \frac{d}{dx}(f(x)+g(x)) \\ &= \frac{d}{dx}f(x) + \frac{d}{dx}g(x) \\ &= f'(x) + g'(x) \\ D(f+g) &= D(f) + D(g) \end{aligned}$$

$$\begin{aligned} D(kf) &= (kf(x))' = \frac{d}{dx}(kf(x)) \\ &= k \frac{d}{dx}(f(x)) \\ &= kD(f) \end{aligned}$$

TRY THE FOLLOWING

Let $V = \overline{C[0, 1]}$ (continuous functions from 0 to 1)

Let $J: V \rightarrow \mathbb{R}$ be defined by
 ↗ real number space.

$$J(f) = \int_0^1 f(x) dx$$

Prove that J is linear transformation from V to \mathbb{R} .

SOLUTION:

Let $f, g \in V$

$$J(f+g) = \int_0^1 (f+g)(x) dx$$

$$= \int_0^1 (f(x) + g(x)) dx$$

$$= \int_0^1 f(x) dx + \int_0^1 g(x) dx$$

$$J(f+g) = J(f) + J(g)$$

$$\text{also } J(kf) = \int_0^1 kf(x) dx = k \int_0^1 f(x) dx$$

$$J(kf) = k J(f)$$

No:

Q) THE FOLLOWING

$R^n \rightarrow R^m$ given by

$$T(x) = Ax = b$$

$A \rightarrow m \times n$ matrix

$x \rightarrow x \rightarrow n \times 1$ matrix (column vector)

$b \rightarrow m \times 1$ matrix (column vector)

Check whether T is linear? $\rightarrow n \times 1$

NOTE $Ax = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & & \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = b \rightarrow m \times 1$

$$b = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix} \quad x \in R^n \quad b \in R^m$$

$$T: R^n \rightarrow R^m$$

Let $x_1, x_2 \in R^n$

$$\begin{aligned} T(x_1 + x_2) &= A(x_1 + x_2) \\ &= Ax_1 + Ax_2 = T(x_1) + T(x_2) \end{aligned}$$

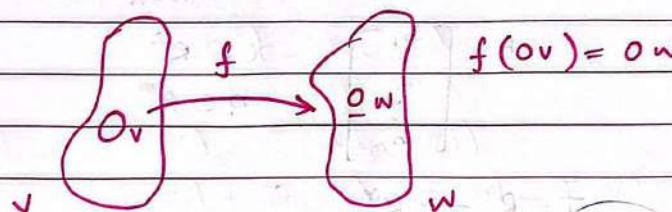
$$\begin{aligned} \text{also } T(kx_1) &= A(kx_1) = kAx_1 \\ &= kT(x_1) \end{aligned}$$

$\therefore T(x) = Ax$ is linear

→ $T(x) = Ax$ is linear and is also called MATRIX TRANSFORMATION or a LINEAR transformation called multiplication by A. Here A in $T(x) = Ax$ is called MATRIX OF LINEAR TRANSFORMATION.

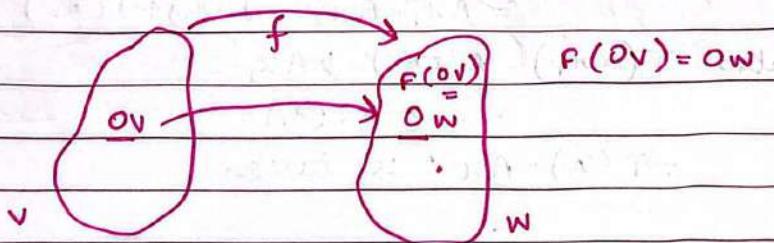
LECTURE 21

If f is a linear transformation from $V \rightarrow W$, then the zero vector of the vector space V is always going to map as zero vector of the vector space W as shown below:



$0v \rightarrow$ zero vector of V

$0w \rightarrow$ zero vector of W



Note: In future we shall use 0 instead of
0v or 0w

EXAMPLES :

$$\textcircled{1} \quad T: R^n \rightarrow R^m$$

$$T(x) = Ax \quad \text{is linear}$$

$$T(0) = A0 = 0$$

$$\Rightarrow T(0) = 0$$

We already proved the following transformations as linear.

$$\textcircled{2} \quad F \left(\begin{bmatrix} x \\ y \end{bmatrix} \right) = \begin{bmatrix} x-y \\ x+y \\ 5x \end{bmatrix}$$

it is seen that

$$F \left(\begin{bmatrix} 0 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \rightarrow \text{zero of } R^3$$

↓ zero of
 R^2

$$\textcircled{3} \quad J: V \rightarrow R \quad V = C[0, 1]$$

$$J(f) = \int_0^1 f(x) dx$$

$$J(0) = \int_0^1 0 dx = 0$$

→ mapping vector into $\mathbf{0}$ (kernel / nullspace)

day / date:

(4) $D: W \rightarrow V$

$$D(f) = f'(x) = \frac{d}{dx} (f(x))$$

$$D(\mathbf{0}) = \frac{d}{dx} (\mathbf{0}) = \mathbf{0}$$

but how to prove in general?

X

PROOF

If $T: V \rightarrow W$ is a linear transformation, then
 $T(\mathbf{0}) = \mathbf{0}$.

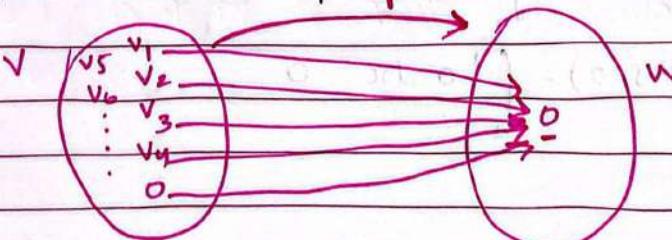
PROOF

$$\begin{array}{ccc} & \xrightarrow{\text{vector}} & \xrightarrow{\text{scalar}} \\ T(\mathbf{0}) & = T(0v) & = 0T(v) = \mathbf{0} \\ v \in V & \quad \because T(kv) = kT(v) \end{array}$$

OR for any $v \in V$

$$\begin{aligned} T(\mathbf{0}) &= T(v-v) = T(v+(-v)) \\ &= T(v) + T(-v) = T(v) - T(v) = \mathbf{0} \end{aligned}$$

→ If $T: V \rightarrow W$ is a linear transformation, then the set of vectors in V that maps into $\mathbf{0}$ is called the **kernel** (or nullspace) of T ; it is defined denoted by $\text{KER}(T)$.





$$\text{KER}(T) = \{v_1, v_2, v_3, v_4, 0\}$$

$$T(v_1) = T(v_2) = T(v_3) = T(v_4) = T(0) = 0$$

EXAMPLES

$$\textcircled{1} \quad D: V \rightarrow W$$

$$\Rightarrow D(f) = f'(x)$$

$\text{KER}(D) = \text{set of all functions s.t. } D(f) = 0$

$$\Rightarrow f'(x) = 0 \Rightarrow f(x) = k$$

$k \rightarrow \text{constant}$

$\therefore \text{KER}(D) = \text{set of all constant functions}$
in V . *differentiating a constant gives you 0.

$$\textcircled{2} \quad J: P_1 \rightarrow \mathbb{R} \quad P_1(x) \quad \left. \begin{array}{l} \text{polynomials} \\ \text{of degree} = 1 \end{array} \right\}$$

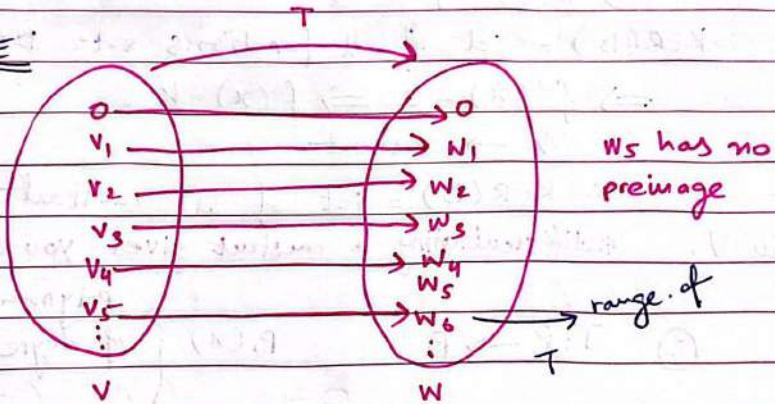
Find $\text{ker}(J)$, where $J(P) = \int_0^1 P(x) dx$

ANSWER $\text{KER}(J) = \text{constants of all polynomials}$
of form $P(x) = kx$
 $k \rightarrow \text{constant}$

RANGE OF LINEAR TRANSFORMATION

If $T: V \rightarrow W$ is linear then set of all vectors in W that are images under T of atleast one vector in V is called range of T . It is denoted by $R(T)$

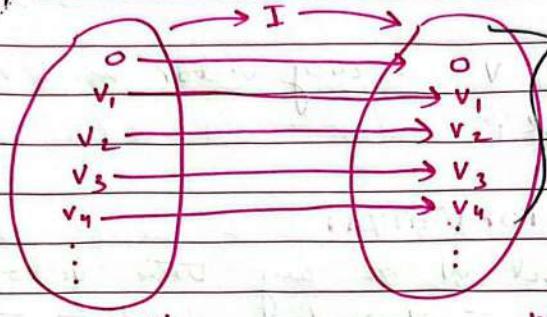
EXAMPLE



$$R(T) = \{0, w_1, w_2, w_3, w_4, w_6, w_7, \dots\}$$

IDENTITY TRANSFORMATION

Let V be any vector space the mapping $I: V \rightarrow V$ defined determined by $I(v) = v$ is called IDENTITY OPERATOR as shown in the figure.



$$\Rightarrow I(v_1) = v'_1, I(v_2) = v'_2, \dots, I(0) = 0 \text{ etc.}$$

TRY THE FOLLOWING

If $I: V \rightarrow W$ is an identity transformation
ie $I(v) = v \quad \forall v \in V$ then

① I is linear

② find $R(I)$, ③ $\text{KER}(I)$

SOLUTION

$$1) \text{ let } u, v \in V \Rightarrow ku \in V = u + v \in V$$

$$\Rightarrow I(u+v) = u+v = I(u) + I(v)$$

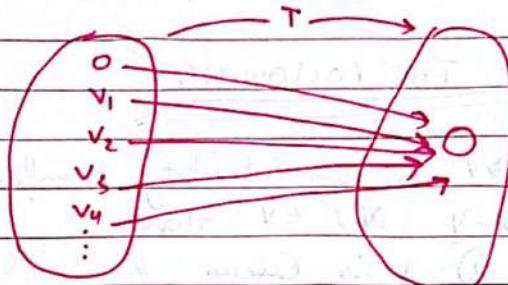
also $I(ku) = ku \Rightarrow kI(u)$

3) $\text{KER}(I) = 0 \because 0 \text{ is the only vector which maps into } 0.$

2) $R(T) = V$, i.e. every vector in V has a preimage.

ZERO TRANSFORMATION

Let V and W be any two vector spaces. The mapping $T: V \rightarrow W$ such that $T(v) = 0$ for every v in V is called zero transformation.



$$\rightarrow T(v_1) = T(v_2) = \dots = 0$$

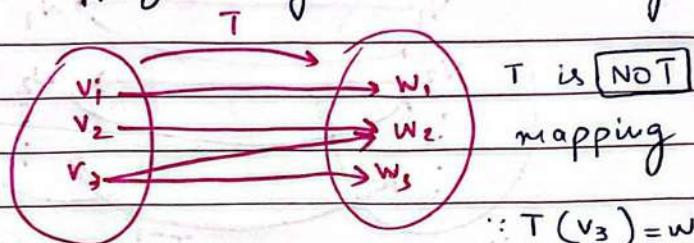
→ A linear transformation $T: V \rightarrow W$ is said to be ONE-TO-ONE if T maps distinct vectors in V into distinct vectors in W .

EXAMPLES

IDENTITY TRANSFORMATION is one-to-one, & rev
 $I(v) = v$, $I: V \rightarrow V$

NOTE: **ZERO TRANSFORMATION** is not one-to-one

NOTE: In a mapping every element has only one image

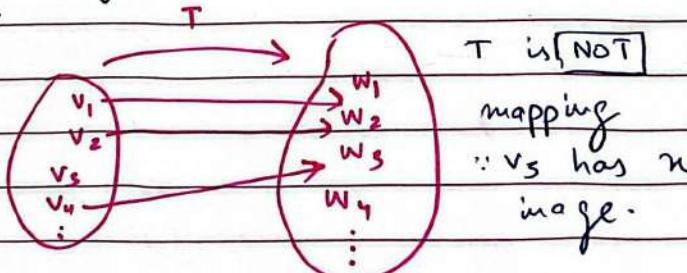


$\therefore T(v_3) = w_2$ and

$T(v_3) = w_3$

v_3 has two images

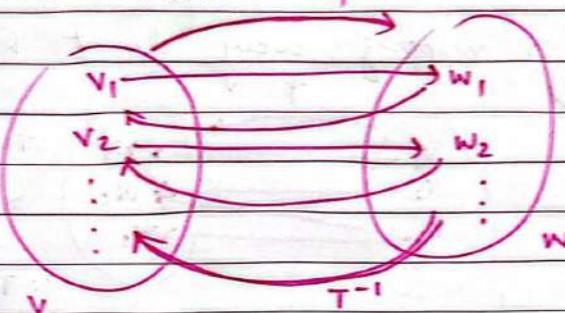
In addition every element in domain must have an image.



INVERSE LINEAR TRANSFORMATION.

If $T: V \rightarrow W$ is linear and one-to-one then the INVERSE LINEAR TRANSFORMATION is given by

$T^{-1}: R(T) \rightarrow V$ which maps $w \in R(T)$ back into $v \in V$



$\therefore T$ is one-to-one \Rightarrow each vector in $R(T)$ is image of a unique vector in V . $R(T)$ may or may not be all of W .