

1) $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$ Dimension=2 $\in R^2$ $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ Dimension=3 $\in R^3$

2) $\begin{bmatrix} 8 \\ 13 \end{bmatrix} = 3 \begin{bmatrix} 2 \\ 3 \end{bmatrix} + 2 \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

↳ linear combination of $\begin{bmatrix} 2 \\ 3 \end{bmatrix}$ and $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$

3) Linear dependency is asked for a set of vector(s).

4) If at least one vector of set can be represented as linear combination of others then set is linearly dependent.

5) If $w = 3u + 0v$ then v cannot be represented in terms of u and w using these coefficient. ^{there may be other coeff}

6) A set containing zero vector is always LD.

7) If subset $\{u_1, u_2\}$ is LD then superset $\{u_1, u_2, u_3\}$ is also LD.

8) $c_1v_1 + c_2v_2 + c_3v_3 + \dots + c_nv_n = 0$
(Do you have at least one $c_i \neq 0$)

YES → LD (Non trivial solⁿ also exist along with trivial solⁿ)
NO → LI (Only one trivial solⁿ, all $c_i = 0$) ^{obvious}

1) Vector space: collection of vectors (must have a zero vector) Ex: R^2, R^3

2) Filling the vector space (span):
Any ' n ' LI vectors fills space of R^n .
It means, any vector in R^n can be represented as linear combination of these ' n ' LI vectors.

3) $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$ is LI so it fills space of R^2 .
It's convenient vectors becoz getting coefficient is so easy
 $\begin{bmatrix} 4 \\ -2 \end{bmatrix} = 4 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + (-2) \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

4) $\left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 3 \\ 6 \end{bmatrix} \right\}$ is LD so doesn't fill space of R^2 .
Some vectors in R^2 may be represented as linear combination of these two vector, but not all.

5) $a \begin{bmatrix} 1 \\ 2 \end{bmatrix} + b \begin{bmatrix} 9 \\ 2 \end{bmatrix}$ is also a vector space filled by these 2 vectors.

6) Basis: Most efficient vector set that can span the space.

Ex: $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 3 \\ 3 \end{bmatrix} \right\}$ There are 2 LI vectors. It fills space of R^2 .

Basis is $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$

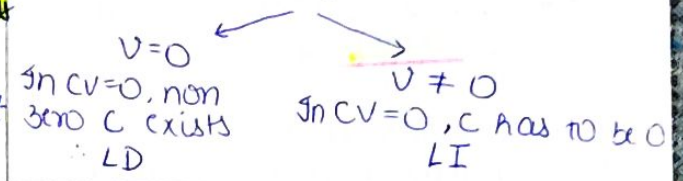
1) We can not have more than ' n ' LI vectors in R^n .

Use these intuitions to test LD/LI:

- For 2 vectors check if 1 is multiple of other.
- If set has zero vector \Rightarrow LD
- If set has more than n vectors $\in R^n \Rightarrow$ LD

8) In a matrix 3×5 , there are 5 vectors of R^3 .

9) If there is just one vector in set $\{v\}$



1) Matrix x Vector

$$\begin{bmatrix} 2 & 4 \\ 3 & 5 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = a \begin{bmatrix} 2 \\ 3 \end{bmatrix} + b \begin{bmatrix} 4 \\ 5 \end{bmatrix}$$

Linear combination of columns of matrix.

2) In $Ax=b$, b is linear comb. of columns of A and coefficient is from vector x .

3) $\left\{ \begin{bmatrix} 5 \\ 0 \end{bmatrix}, \begin{bmatrix} 10 \\ 1 \end{bmatrix} \right\}$ cannot be produced using 0 so it's LI

4) No. of vectors in basis of $R^n = n$

System of Linear eqⁿ

1) Linear eqⁿ: $Ax + By = c$

2) Non linear eqⁿ: $x^2 + y = 5$,
 $x - xy = 4$, $\sqrt{x} + y = 2$



4) $Ax=0$ is always consistent as it has trivial solⁿ $x=0$.

$$a_1x_1 + a_2x_2 + \dots = 0$$

Is there any non zero x_i (Non trivial solⁿ)

columns of A are LD \Rightarrow LI

- 5) Solⁿ exist or not for $Ax=b$?
- If column vectors of A fills space then solⁿ always exist.
 - Otherwise it depends on b
 - if b is linear comb. of columns of A then solⁿ exist
 - else no.

Multiplying 2 matrices :-

1) $[AB]_{12} = (\text{1st row of } A) \cdot (\text{2nd column of } B)$

2) $\begin{bmatrix} 2 & 3 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 1[2]+0[3] & 2[2]+1[3] \\ 1[1]+0[0] & 2[1]+1[0] \end{bmatrix}$

A B

3) $[A][B] = [C]$

If there are 2 LI vectors in A ; C can have at most 2 LI vectors as it's the linear comb. of columns of A .

4) If system has m eqⁿ with n variables then matrix formed is $m \times n$.

5) $\begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix} \begin{bmatrix} 2 & 1 & 0 \end{bmatrix}$ This is a way to generate rank 1 matrix (ie no. of LI vectors = 1)

6) $\begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 3 & 6 & 9 \end{bmatrix}$ col 2 & col 3 are redundant as it can be generated using col 1.

Solⁿ of $Ax=b$

$A_{m \times n}$ has m LI vectors (columns)?

Yes: Solⁿ always exist
No: b is linear comb. of columns of A ?

Yes: Solⁿ exist
No: NO solⁿ

b is linear comb. of col^s of A ?

Yes: Always a solⁿ
No: Never a solⁿ

Yes: unique solⁿ
No: infinite solⁿ

1) Column space: linear combination of columns.

2) Echelon form / Row echelon form

- All non zero rows are above zero rows.
- All ele below leading entry in a column are 0.

is not true for $Ax=b$ becoz b also need to be changed accordingly.

• The leading entry of any row occurs to the right of leading entry of row above it.

3) Leading entry: 1st non zero ele of a row.

TRICK: Make stair with leading entries, if its stair with a step in each row then its in echelon form.

4) Pivot column \rightarrow LI col
free column \rightarrow LD col on pivot

5) In echelon form, column having leading entry is called pivot variable / column.

Remaining are free var / column.

Ex: $x + 3y + z = 2$

$0x - 2y + 2z = 5$

$0x + 0y + 0z = 0$

x & y are pivot var / column
 z is free var / column.

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

6) Augmented matrix $[A|b]$ line is optional
 A is coefficient matrix.

7) Don't label b as pivot or free as its not associated with any var.

8) Which are pivot & free in augm. matrix

$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$ don't consider b
Pivot

9) Elementary row operation:-

- $R_i \leftrightarrow R_j$
- $R_i \rightarrow cR_i$, $c \neq 0$
- $R_i \rightarrow R_i + cR_j$

10) Just make sure that no info is lost by our row operation.

11) Gaussian Elimination
Matrix \rightarrow Echelon form

Rank

1) Rank = # linearly independent columns = # linearly independent rows = # non zero rows in echelon form = # pivot variables in echelon form.

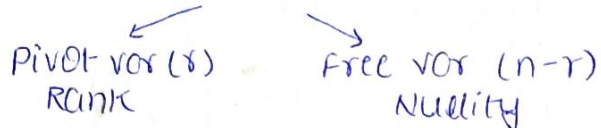
2) Rank ≥ 0

3) Rank is zero only for NULL (zero) matrix.

4) To find rank, convert matrix to echelon form & find rank.

5) Nullity: NO. of free variables.

n columns in coeff matrix



6) Total no. of variables (columns) = RANK + Nullity.

7) System of linear eqⁿs:-

• Homogeneous $Ax=0$

• Non Homogeneous (Heterogeneous) $Ax=b$

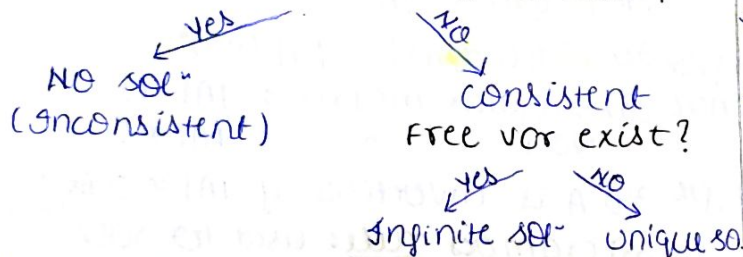
8) To solve system of linear eqⁿs:-

Find Augmented matrix \rightarrow convert A into echelon form \rightarrow form equations and solve.

Number of solⁿs:

Find echelon form of A in $[A|b]$

$[0\ 0\ 0\ 0\ |\ \text{Nonzero}]$ exist?



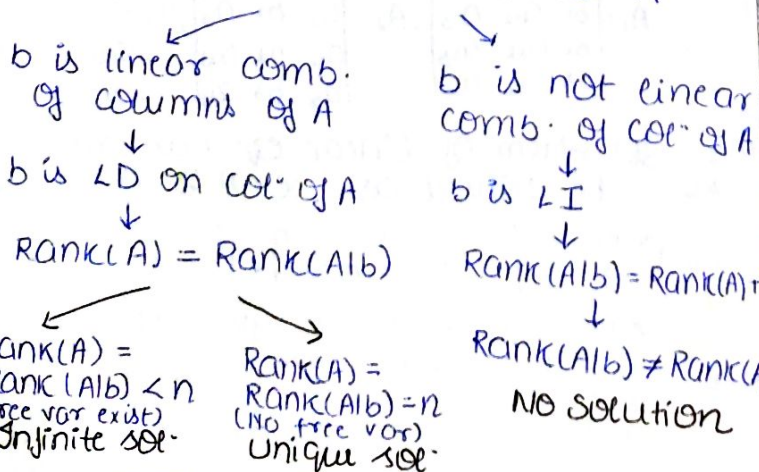
9) For infinite sol:

NO. of LI solⁿ = NO. of free var in A = Nullity of A

*Take free var as some parameter. = NO. of parameters in sol.

10) $\text{Rank}(A_{m \times n}) \leq \min(m, n)$

11) When we add vector b to A



*Get dependency in matrix, echelon, reduced echelon form is same as all obtained via elementary row op.

12) Elementary row operations does not change linear dependency between columns.

$A \xrightarrow{\text{unknown ele.}} B$
 $C_3 = C_1 + 2C_2$ row operation. $C_3 = C_1 + 2C_2$

13) E.C. row operations preserve row space but not column space

14) $a \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} + b \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix}$
 Dimension of vector space = NO. of LI vector = 2
 Dimension of vector = 3

1) Row reduced echelon form or Reduced echelon form

It has these additional properties:

- Each leading entry is 1.
- All ele above & below leading entry is 0.

2) converting into reduced ech. form:

- convert to echelon form.
- Make leading entry 1
- Make all ele above leading entry 0.

3) Adv: Pivot columns are convenient vectors so finding coeffs is easy.

4) Echelon form of a matrix is not unique whereas reduced echelon form is unique.

5) $Ax=0$ has "only trivial sol" means system has unique sol.

Determinant

• Here all matrices are square mat.

1) $\det(A)$ or $|A|$ is scalar.

2) $|I| = 1$

3) \det changes sign when two rows (or two columns) are interchanged

4) Linearity is applied on a row or column at a time.

$$\begin{vmatrix} a+b & c+d \\ e & f \end{vmatrix} = \begin{vmatrix} a & c \\ e & f \end{vmatrix} + \begin{vmatrix} b & d \\ e & f \end{vmatrix}$$

$$\begin{vmatrix} ka & kb \\ c & d \end{vmatrix} = k \begin{vmatrix} a & b \\ c & d \end{vmatrix}$$

5) $|A+B| \neq |A| + |B|$

6) If any 2 rows (or columns) are equal/proportional then \det is 0.

7) $R_i \rightarrow R_i \pm cR_j$

det remains unchanged.

8) whatever you do with rows is also applicable for columns in det.

9) Matrix with row^{or} column of zeroes has det 0.

10) Det. of diagonal/triangular matrix is product of diagonal elements.

11) Finding det using permutation

Take ^{all} perm; such that each row & each column has 1 nonzero ele. then make it into diagonal matrix & take product of diagonal ele.

12) using it we can find sign of terms

Ex: $a_{13} a_{22} a_{34} a_{41}$

$3241 \rightarrow 1243 \rightarrow 1234$

2 swaps $\Rightarrow (-1)^2 \Rightarrow +ve$

13) Calculate determinant :-

• using permutation

• using cofactor

• convert into echelon form (triangular matrix), take product of diagonal elements.

14) Minor M_{ij} : leave i th row and j th column, take determinant.

15) Cofactor $C_{ij} = (-1)^{i+j} M_{ij}$

16) Adjoint : transpose of cofactor element matrix.

17) Determinant is sum of product of any row (or column) ele with its cofactor.

18) C_{ij} do not depend on ele at position (i,j) .

19) If ele of a row or column is multiplied by cofactor of any other row or column, its Σ is 0.

Ex: $a_{21}C_{11} + a_{22}C_{12} + a_{23}C_{13} = 0$

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

21) If A is singular (ie $|A|=0$) then A^{-1} does not exist.

22) $AA^{-1} = A^{-1}A = I$

23) For square matrix (Both A inverse of each other.)
 $AB = I \Leftrightarrow BA = I$

Trick to find adjoint :-

$\begin{bmatrix} a & c \\ b & d \end{bmatrix} \Rightarrow \begin{bmatrix} d & -c \\ -b & a \end{bmatrix}$

$\begin{bmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{bmatrix} \xrightarrow{\text{swap}} \begin{bmatrix} 2 & 5 \\ 3 & 6 \end{bmatrix} \xrightarrow{\text{swap}} \begin{bmatrix} 3 & 6 \\ 2 & 5 \end{bmatrix} \xrightarrow{\text{swap}} \begin{bmatrix} 3 & 6 \\ 2 & 5 \end{bmatrix}$

24) Det. evaluated across any row or column is same.

25) $|A^T| = |A|$

26) $|AB| = |B||A| = |A||B|$

27) $|A^n| = (|A|)^n$

28) $|A^{-1}| = \frac{1}{|A|}$

29) $A \cdot \text{adj}(A) = (\text{adj}(A)) \cdot A = |A| \cdot I$

30) $\text{adj}(AB) = \text{adj}(B) \text{adj}(A)$

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

32) $|\text{adj}(\text{adj}(A))| = |A|^{(n-1)^2}$

33) singular matrix : $|A|=0$
Non " " : $|A| \neq 0$

34) A is invertible if $|A| \neq 0$ ie A^{-1} exist

35) Cramer's rule : used to solve system of linear eq. $TC = O(n^3)$

• condition: $|A| \neq 0$ ie A^{-1} exist ie unique sol. exist

$x = \frac{|A|}{|A|} \begin{bmatrix} |A_1| \\ |A_2| \\ |A_3| \end{bmatrix} \quad \begin{matrix} Ax = b \\ x = A^{-1}b \end{matrix}$

$A_1 = \begin{bmatrix} b_1 & a_{12} & a_{13} \\ b_2 & a_{22} & a_{23} \\ b_3 & a_{32} & a_{33} \end{bmatrix}, A_2 = \begin{bmatrix} a_{11} & b_1 & a_{13} \\ a_{21} & b_2 & a_{23} \\ a_{31} & b_3 & a_{33} \end{bmatrix}, A_3 = \begin{bmatrix} a_{11} & a_{12} & b_1 \\ a_{21} & a_{22} & b_2 \\ a_{31} & a_{32} & b_3 \end{bmatrix}$

36) System of linear eq. can also be solved as $x = A^{-1}b$, using echelon form

* To find 3rd column of A^{-1}
 $A^{-1} \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = x \Rightarrow Ax = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$ solve for x

* Methods to solve system of linear eq.

i) using echelon form

ii) $x = A^{-1}b$

iii) using Cramer's rule

Eigen value & Eigen Vector

• We will talk about **square matrix**.

1) When we multiply a vector with matrix and we get vector in same direction then it's eigen vector.

2) $Ax = \lambda x$ x : Eigen vector
 λ : Eigen value (scalar)

3) $\lambda = 1$ No change
 $\lambda < 1$ Shrink
 $\lambda > 1$ Stretched

4) λ can be 0 but eigen vector can't be zero vector.

5) If x is eigen vector of A then kx ($k \neq 0$) is also eigen vector of A .
Eigen value is same for both (corresponds to each Eigen vector)

6) Every matrix has infinitely many eigen vector, and at most n LI Eigen vector.

7) Calculating Eigen value & Eigen vector:

• Find Eigen values from $|A - \lambda I| = 0$ (char eq)
• Find Eigen vectors from $(A - \lambda I)x = 0$

8) Every square matrix of degree $n \times n$ has n Eigen values. (may be repeated)

9) If matrix is of order $n \times n$ then char eqⁿ is of degree n .

10) $\lambda_1 \neq \lambda_2 \Rightarrow e_1 \& e_2$ are LI

$\lambda_1 = \lambda_2 \Rightarrow$ If λ is repeating twice (λ^2) then we may get at most two LI Eigen vectors.

11) $(\lambda - 1)^2 (\lambda - 5) (\lambda - 7)^3$ Matrix must be of order 6×6
Eigen vectors: at least 3
LI at most 6

12) $(\lambda - \lambda_1)^m$

Algebraic mul.: (m) no. of times λ_1 is repeated.

Geometric multiplicity: NO. of LI Eigen vectors corresponding to λ_1

13) $AM \geq GM$

14) For $A_{n \times n}$, n Eigen values exist and at most n LI Eigen vectors.

* Orthogonal \rightarrow Linearly independent

15) Real symmetric matrix always has n linearly inde. Eigen vectors even if eigen values are repeating.
ie. $AM = GM$

16) Real sym matrix has:

• n Real Eigen values.

• n Orthogonal (LI) Eigen vectors.

17) Eigen vectors of sym. matrix are orthogonal, when we put them in matrix, it's orthogonal matrix

18) Diagonalizable matrix \leftrightarrow has n LI Eigen vectors.

19) Symmetric matrix is diagonalizable

20) Determinant of any matrix is product of eigen values.

21) Trace (sum of principle diagonal el.) = sum of eigen values.

22) If rows/columns of matrix are dependent ie. $|A| = 0$ then at least one eigen value is 0.

Value of $A = 0$?

* If no λ is 0 means $|A| \neq 0 \Rightarrow \text{Rank}(A) = n$

Yes \rightarrow Rank $(A) < n$
No \rightarrow Rank $= n$

No. of LI vectors corresponding to $\lambda = 0$ = Nullity of A .

23) $(A - \lambda I)x = 0$

It always has free var as $|A - \lambda I| = 0 \therefore$ Infinite Eigen vectors $\forall \lambda$.

24) NO. of LI Eigen vectors corresponding ($\lambda = 0$) = Nullity of A .

25) Cayley Hamilton theorem:

Every square matrix satisfies its own characteristics eqⁿ.

• Used to find A^{-1} and A^k

NOTE: Elementary row op do not preserve Eigen values, so never apply row op in order to find Eigen values

26) AB

BA

Eigen value: λ

λ

($\lambda \neq 0$)

Eigen vector: x

Bx

27) AB and BA shares non-zero Eigen values. (copy non-zero Eigen values & fill remaining with 0's)

Nonzero values of $A^T B$ & BA^T same

- 28) If $A_{4 \times 3}$, $B_{3 \times 4}$ then there can be atmost 3 nonzero values in AB & BA as AB & BA shares nonzero values.
- 29) If value of A is λ then value of A^k is λ^k .
(Evector same 'x' in point 29, 30, 31)
- 30) If value of A is λ then value of $A^k + 3A^{k-1} + I$ is $(\lambda^k + 3\lambda^{k-1} + 1)$.
- 31) If value of A is λ then value of A^{-1} is $1/\lambda$.
- 32) If $(a+bi)$ is value of A then $(a-bi)$ is also an value of A .
- 33) If A is a column vector then $A^T A = |A|^2$
If $A = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$, $|A| = \sqrt{1^2 + 2^2 + 3^2}$ Magnitude/length or Euclidean norm

NOTE: If Evector is of dimension 2×1 then matrix A is of 2×2

- 34) If $\lambda = 0 \Rightarrow |A| = 0 \Rightarrow A^{-1}$ does not exist
If $\lambda \neq 0 \Rightarrow |A| \neq 0 \Rightarrow A^{-1}$ exist.

\Rightarrow LU Decomposition:

Matrix 'A' can be decomposed only if A can be reduced to echelon form without row interchange.

$$A = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{bmatrix} \begin{bmatrix} U_{11} & U_{12} & U_{13} \\ 0 & U_{22} & U_{23} \\ 0 & 0 & U_{33} \end{bmatrix} \Rightarrow A = LU$$

L U

we may have 1's in diagonals of L or U.

Method 1: Make eqⁿ from above & solve.

Method 2: $A = \begin{bmatrix} 1 & 4 & -3 \\ -2 & 8 & 5 \\ 3 & 4 & 7 \end{bmatrix}$ $L = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 3 & - & 1 \end{bmatrix}$

Convert A to echelon form

$R_2 \rightarrow R_2 + 2R_1$
it becomes $\begin{bmatrix} 1 & 4 & -3 \\ 0 & 16 & -1 \\ -2 & 4 & 7 \end{bmatrix}$
0 so add -2 into L

similarly, keep filling L and A will be reduced to U.

*** Solving system of linear eqⁿ $Ax = b$**

- Decompose $A = LU$ then $LUX = b$
- Let $UX = y$ then $Ly = b$
Solve it to get y .
- Solve $UX = y$ to get x .

NOTE: Eigen values of diagonal/triangular matrix are diagonal elements.

Types of Matrices:

1) Identity matrix (I): Square matrix with
 $a_{ij} = \begin{cases} 1 & i=j \\ 0 & \text{else} \end{cases}$

*** Inverse:**

- $A^{-1}A = AA^{-1} = I$
- $(A^{-1})^{-1} = A$
- $(kA)^{-1} = \frac{1}{k} A^{-1}$, $k \neq 0$
- $(AB)^{-1} = B^{-1}A^{-1}$
- For square matrix $AB = BA \Leftrightarrow BA = I$

NOTE: AB can be zero even though none of A, B is zero. (when $|A| = 0$)

*** Transpose:** $a_{ij} \leftrightarrow a_{ji}$

- $(A^T)^T = A$
- $(A+B)^T = A^T + B^T$
- $(AB)^T = B^T A^T$
- $(CA)^T = C^T A^T$
- $(A^T)^{-1} = (A^{-1})^T$

2) Lower Δ or matrix: all elements above diagonal are 0.

Upper Δ or matrix: all elements below diagonal are 0.

3) Diagonal matrix: need not be sq
 $a_{ij} = \begin{cases} 0 & i \neq j \\ \text{otherwise anything} \end{cases}$

4) Symmetric matrix: (sq matrix)
 $A^T = A$

5) Skew symmetric matrix: (sq)
 $A^T = -A$
• Diagonal elements are 0.

6) For any sq matrix $A^T A$ and $A A^T$ are symmetric.

7) Any sq real matrix can be represented as sum of symmetric & skew sym matrix
 $A = \underbrace{\frac{1}{2}(A+A^T)}_{\text{Symmetric}} + \underbrace{\frac{1}{2}(A-A^T)}_{\text{skew symmetric}}$

8) Orthogonal vectors: (\perp vector)
 $A \cdot B = |A||B|\cos 90^\circ = |A||B|\cos 90^\circ = 0$

9) Orthonormal vectors: Orthogonal vectors having unit length.

10) Orthogonal matrix: sq matrix whose rows & columns are orthogonal vectors.

- $A^T A = A A^T = I$

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

11) Idempotent matrix: $A^2 = A$ (sq)

- $A^n = A \quad \forall n > 0$

NOTE: $(a^b)^c = (a^c)^b = a^{bc}$
 $a^{b^c} = a^{(b^c)}$

12) Involutory matrix: $A^2 = I$ (sq)

- $A^{-1} = A$

- If A is involutory then A^k is involutory.

13) Nilpotent matrix: sq matrix such that $\exists k$ such that $A^k = 0$

- min value of k is called index.

14) Hermitian matrix: $A^0 = A$ (sq)

* Transpose = Transpose & conjugate (A^0 or A^* or A^H)

- $A^0 = (\bar{A})^T = (A^T)^*$ $|A^0| = |A|$

- $(A^0)^0 = A$

- $A^0 = A^T$ for real matrix

- $(A+B)^0 = A^0 + B^0$

- $(AB)^0 = B^0 A^0$

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

- $(cA)^0 = \bar{c} A^0$, c is complex no.

- Diagonal ele are purely real nos.

15) Skew Hermitian matrix: $A^0 = -A$

- Diagonal ele are purely imaginary or 0.

16) Unitary matrix: $A A^0 = I$

NOTE: If A is symm matrix then all its eigen vectors are orthogonal. **False**

- All LI Evectors are orthogonal.

- If x is EV then kx is also EV but both are not orthogonal.

Q: NO of matrix idempotent & invertible
 $A^2 = A \Rightarrow A = A A^{-1} \Rightarrow A = I$ only I matrix.

* $A_{3 \times 2}$ & $B_{2 \times 3}$ then $|AB|$ must be 0
 ↳ at most 2 LI vectors so $AB_{3 \times 3}$ can have at most 2 LI $\Rightarrow |AB| = 0$

Q: can a symmetric matrix have $A^2 = -I$
 $A^2 + I = 0 \Rightarrow \lambda^2 + 1 = 0 \Rightarrow \lambda = \pm i$
 But sym matrix has real eigen values
 \therefore Not possible.

NOTE: $A^{2024} B^{2023} v$

In such ques just check v is Eigen vector of matrix A & B with Eigen value λ_1 & λ_2 respectively.

$\Rightarrow \lambda_1^{2024} \lambda_2^{2023} v$