

# ENGINEERING

# MATHS

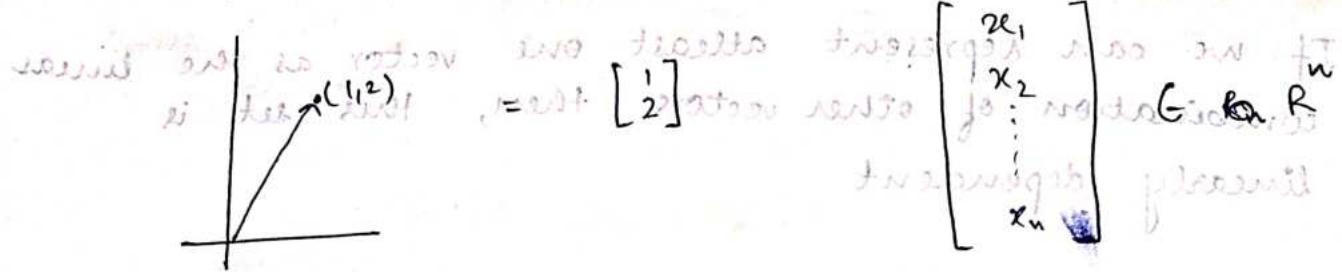
# LINEAR

# ALGEBRA

- system of linear equations
- eigen values & eigen vectors
- determinants.

(1-2 questions)  
(2-3 marks)

Linear algebra helps in data representation



↳ Linear combination of vectors

↳ Multiplying vectors by scalars and adding them together:

$$x_1 A_1 + x_2 A_2 + x_3 A_3 = \underline{\quad}$$

↳ Linear combination of  $A_1 A_2 A_3$

↳ Linearly Dependent vectors -

$\begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \end{bmatrix}$  are linearly dependent [one is scalar multiple of other]

↳ Property is for set of vectors

① 3 vectors are linearly dependent if one of them is linear combination of the other two.

$\rightarrow \left( \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \\ 7 \end{bmatrix}, \begin{bmatrix} 3 \\ 6 \\ 10 \end{bmatrix} \right)$  are linearly dependent

$$\begin{bmatrix} 3 \\ 6 \\ 10 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} + 1 \begin{bmatrix} 2 \\ 4 \\ 7 \end{bmatrix}$$

$\rightarrow \left( \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \\ 7 \end{bmatrix}, \begin{bmatrix} 3 \\ 6 \\ 9 \end{bmatrix} \right)$  linearly dependent

↳  $\begin{bmatrix} 3 \\ 6 \\ 9 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} + 0 \begin{bmatrix} 2 \\ 4 \\ 7 \end{bmatrix}$

$$\begin{bmatrix} 3 \\ 6 \\ 9 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} + 0 \begin{bmatrix} 2 \\ 4 \\ 7 \end{bmatrix}$$

$\{u, v, w, p, q, r\} \rightarrow$  set of vectors

If we can represent atleast one vector as the linear combination of other vectors, then, this set is linearly dependent.



① A set containing zero vector is always linearly dependent.

set

$$u = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

$$v = \begin{bmatrix} 2 \\ 4 \\ 9 \end{bmatrix}$$

$$w = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

zero vector

exists in set

$$= c_1 u + c_2 v + c_3 w$$

$$w = 0 \cdot u + 0 \cdot v$$

$\therefore$  linearly dependent.

→ vector linearly dependent  
exists in set

$$v_1 = c_2 v_2 + c_3 v_3 + c_4 v_4 + \dots + c_n v_n$$

Can  $v_2$  be represented as linear combination of  
rest of other vectors?

out Ans - Yes only if  $c_2 \neq 0$ , then no  
if  $c_2 = 0$ , then yes

present in  $(\begin{bmatrix} 8 \\ 3 \\ 0 \end{bmatrix}, \begin{bmatrix} 5 \\ 7 \\ 1 \end{bmatrix}, \begin{bmatrix} 7 \\ 2 \\ 8 \end{bmatrix})$

Let  $u_i$ 's be the vectors in  $\mathbb{R}^n$  for  $i = 1, 2, 3, 4$

Which of the following  $\{u\}$  are correct?

A. If  $\{u_1, u_2, u_3\}$  is linearly independent, so is  $\{u_1, u_2, u_3, u_4\}$

B. If  $u_4$  is not linear combination of  $\{u_1, u_2, u_3\}$   
then  $\{u_1, u_2, u_3, u_4\}$  linearly independent

C. Any set containing zero vector is linearly dependent

D. If  $\{u_1, u_2, u_3\}$  is linearly dependent, so is  $\{u_1, u_2, u_3, u_4\}$

\* If a subset is linearly dependent, then, its superset is also linearly dependent.

### Linear independence

a set of vectors is linearly independent iff they are not linearly dependent.

If  $c_1v_1 + c_2v_2 + c_3v_3 + \dots + c_nv_n = 0$ , then

If any one of  $c_1, c_2, c_3, \dots, c_n$  is non-zero, then, it is linearly dependent set

Suppose  $c_1 \neq 0$

then,  $v_1 = -\frac{1}{c_1}(c_2v_2 + c_3v_3 + \dots + c_nv_n)$

may or may not be  
else linearly independent

Convenient vectors in  $\mathbb{R}^2$

$\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}$  are convenient vectors

for reason  $\begin{bmatrix} x \\ y \end{bmatrix} = \alpha \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \beta \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

any vector can be expressed as linear combination of these two vectors

If there are 2 linearly independent vectors,

then, any other vector in the vector space  
can be derived from these two

\* If 2 vectors in  $\mathbb{R}^2$  are linearly independent, then, any set containing those 2 vectors are linearly dependent.

Can we have more than  $i$  independent vectors in  $\mathbb{R}^i$

$\rightarrow$  No

\* If there are more than  $n$  vectors in  $\mathbb{R}^n$  in the set, then, the set is definitely linearly dependent.

(Now, now is not so if we take two for which the given set is not)

Multiplying a matrix by a vector

$$\begin{bmatrix} a & d & g \\ b & e & h \\ c & f & i \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} ax + dy + gz \\ bx + ey + fz \\ cx + fy + iz \end{bmatrix}$$

$$= x \begin{bmatrix} a \\ b \\ c \end{bmatrix} + y \begin{bmatrix} d \\ e \\ f \end{bmatrix} + z \begin{bmatrix} g \\ h \\ i \end{bmatrix} \quad (\text{in vector form})$$

linear result is linear combination of column vectors of the matrix where coefficients are from vector  $x$

$AX = B$

matrix    vector    vector

\* If  $AX=0$  has some non trivial solution then, columns of  $A$  are linearly dependent.

## System of linear equations

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

} system of linear equations

linear combination of columns  $\rightarrow$  towards

$$\begin{aligned} x - 2y &= 1 \\ 3x + 2y &= 11 \end{aligned}$$

method II S

incongruous system

$$\text{coefficient matrix: } \begin{bmatrix} 1 & -2 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 1 \\ 11 \end{bmatrix} = x \begin{bmatrix} 1 \\ 3 \end{bmatrix} + y \begin{bmatrix} -2 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 11 \end{bmatrix}$$

matrix

$$AX = B$$

$Ax = 0$  has some non trivial solution then columns of  $A$  are linearly dependent? True/False

$$\begin{bmatrix} a_1 & a_2 & a_3 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix}$$

linear combination of column vector

$$\text{if } c_1a_1 + c_2a_2 + c_3a_3 = 0$$

then  $c_1, c_2, c_3$  are non zero

at least one of  $c_1, c_2, c_3$  is non zero

$$\therefore a_1 = -\frac{1}{c_1}(c_2a_2 + c_3a_3)$$

$A$  for which  $a_1$  is not zero

not linearly independent

more equations  
than variables

single variable &  
more than one

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a \begin{bmatrix} 1 \\ 3 \end{bmatrix} + b \begin{bmatrix} 2 \\ 4 \end{bmatrix} \\ c \begin{bmatrix} 1 \\ 3 \end{bmatrix} + d \begin{bmatrix} 2 \\ 4 \end{bmatrix} \end{bmatrix}$$

A handwritten diagram labeled 'A' at the top. It consists of a series of five circles arranged horizontally, with a wavy line underneath them. The first circle is enclosed in a bracket on the left side.

2 LI vectors

$$\begin{bmatrix} \text{sd} = \text{N} \times \text{SD} \\ - \end{bmatrix} = \begin{bmatrix} \text{C} \\ - \end{bmatrix}$$

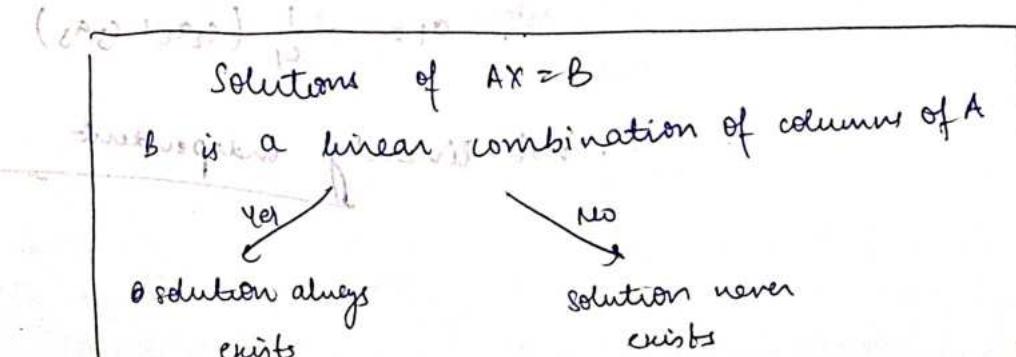
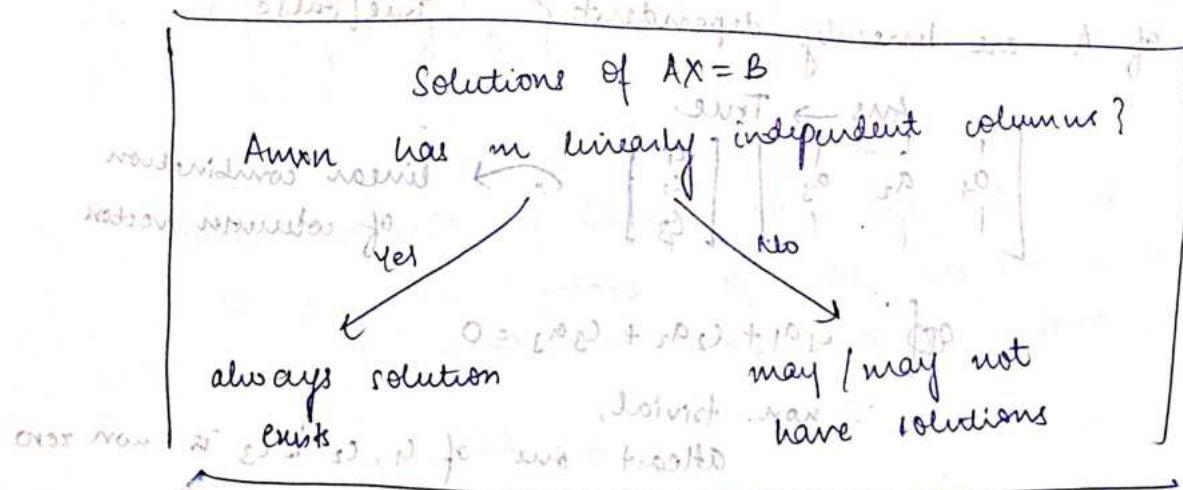
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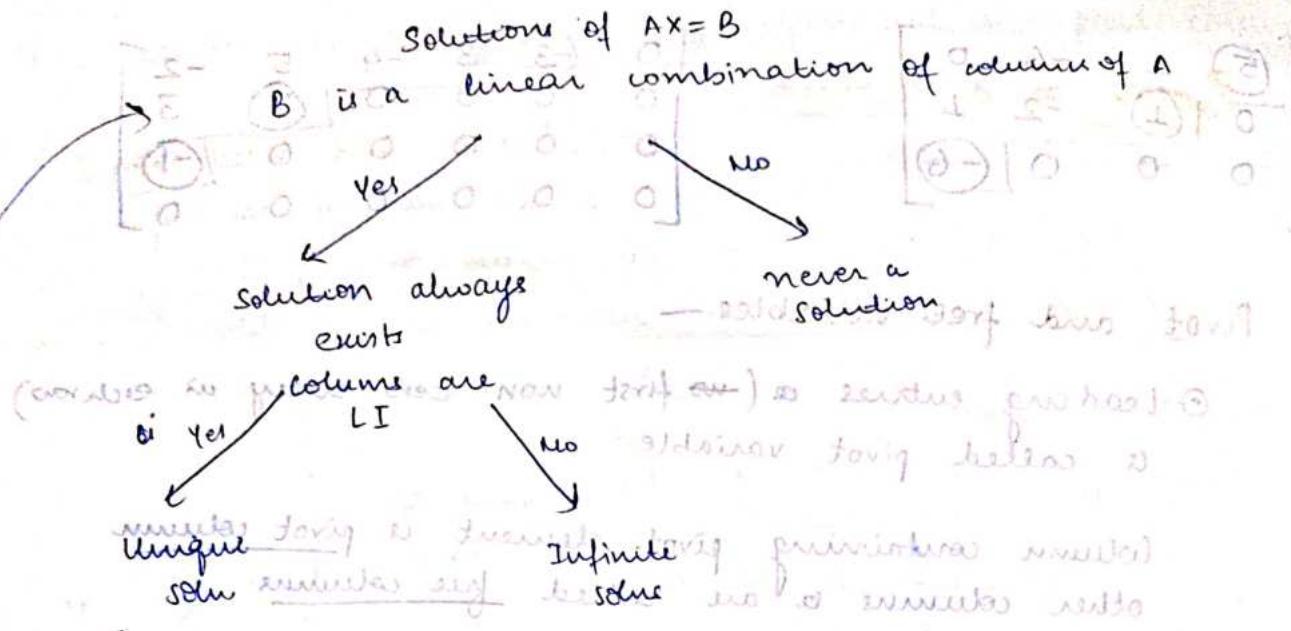
$$\begin{aligned} \text{L1} &= b_1 - \frac{b_2}{3} \\ \text{L2} &= b_2 + \frac{b_1}{3} \end{aligned} \quad \left. \begin{array}{l} \text{for consistency} \\ \text{almost 2} \end{array} \right\} \text{dissolve}$$

$\begin{bmatrix} 1 \\ 3 \end{bmatrix} x = \begin{bmatrix} 1 \\ 11 \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix} \begin{bmatrix} 2 & 5 \\ 5 & 2 \end{bmatrix}$

$\begin{bmatrix} 1 \\ 1 \end{bmatrix}$  linearly independent vectors

\* If  $B$  is not a linear combination of columns of  $A$ , then, we can never get the solution of  $AX = B$ .





If a vector ( $B$ ) can be represented as linear combinations of a few vectors and those vectors are linearly independent, then there is a unique solution.

### Step by step solution of $AX = B$ (Gaussian elimination)

Gaussian elimination - An algorithm to solve the systems of linear equations

Matrix	Gaussian elimination	Echelon form of matrix
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Terms

- Echelon form of matrix
- Pivot and free variable
- Elementary row operations

leading entry		Echelon form -
(a)	d	$\begin{matrix} 1 & 0 & \dots & 0 & 0 \\ 0 & 1 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & 0 \end{matrix}$
(b)	f	
0	0	
0	0	

- ① All non zero rows are above any rows of all zeros
- ② All entries in a column below leading entry are zero
- ③ Leading entry of any row occurs to the right of the leading entry of the row above it.

leading entry → first non-zero entry in each row.

$$\left[ \begin{array}{ccccc} 5 & 1 & -6 & 0 \\ 0 & 1 & -2 & 1 \\ 0 & 0 & 0 & -6 \end{array} \right] \xrightarrow{\text{order switch}} \left[ \begin{array}{ccccc} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

Pivot and free variables -

④ Leading entries ( $\neq$  first non-zero entry in each row) is called pivot variable.

Column containing pivot element is pivot column  
other columns are called free columns. w32

Identify basic and free variables in given augmented matrices —

Since augmented matrix  
is given

∴ last column is neither

more ~~very~~ basic: ~~not?~~

most free note basic get

Diagram illustrating the steps of Gaussian elimination:

- Step 1:** Swap Row 1 and Row 2.
- Step 2:** Add -Row 1 to Row 2.
- Step 3:** Add -Row 1 to Row 3.
- Step 4:** Swap Row 2 and Row 3.
- Step 5:** Add Row 2 to Row 3.

The resulting matrix is:

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

Annotations:

- "pivot column" is written vertically along the first column.
- "pivot element" is written vertically along the main diagonal.
- "zero entries" are labeled in the off-diagonal positions.
- "row operations" are labeled above the rows.
- "column operations" are labeled to the left of the columns.
- "swap rows" is labeled near the swap operation.
- "add multiples" is labeled near the row reduction operations.
- "zero entries" are also labeled near the bottom-right corner.

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

\* Free column is always linearly dependent on pivot columns.

enters in the row

### Elementary Row Operations

- Swap positions of 2 rows  $R_i \leftrightarrow R_j$
- Multiply a row by non zero scalar  $R_i \rightarrow cR_i$
- Add to one row scalar multiple of other  $R_i \rightarrow R_i + cR_j$

Q = identity goes to zero?

Gaussian elimination

$$\left[ \begin{array}{cccc} 1 & 3 & 1 & 9 \\ 1 & 1 & -1 & 1 \\ 3 & 11 & 5 & 35 \end{array} \right] \xrightarrow{\begin{array}{l} R_2 \rightarrow R_2 - R_1 \\ R_3 \rightarrow R_3 - 3R_1 \end{array}} \left[ \begin{array}{cccc} 1 & 3 & 1 & 9 \\ 0 & -2 & -2 & -8 \\ 0 & 8 & 2 & 8 \end{array} \right]$$

$$\xrightarrow{\begin{array}{l} R_2 \rightarrow -R_2 \\ R_3 \rightarrow R_3 + R_2 \end{array}} \left[ \begin{array}{cccc} 1 & 3 & 1 & 9 \\ 0 & 2 & 0 & -8 \\ 0 & 0 & 2 & 0 \end{array} \right]$$

row echelon form

$$\left[ \begin{array}{ccc} 0 & 0 & -3 \\ 9 & 3 & 5 \\ 3 & 1 & 1 \end{array} \right] \xrightarrow{R_1 \leftrightarrow R_3} \left[ \begin{array}{ccc} 3 & 1 & 1 \\ 9 & 3 & 5 \\ 0 & 0 & -3 \end{array} \right] \xrightarrow{R_2 \rightarrow R_2 - 3R_1} \left[ \begin{array}{ccc} 3 & 1 & 1 \\ 0 & 0 & 2 \\ 0 & 0 & -3 \end{array} \right]$$

$$\xrightarrow{R_3 \rightarrow R_3 + \frac{3}{2}R_2} \left[ \begin{array}{ccc} 3 & 1 & 1 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{array} \right]$$

row echelon form

row echelon form

row echelon form

all

nonzero  
entries

row echelon form

examples today we have discussed different types of matrices

## Rank of a matrix

- ↳ linearly independent rows/ columns
- ↳ pivot elements in row echelon form of matrix
- ↳ non zero rows of an echelon form of matrix

$$\boxed{\text{Rank of zero matrix} = 0}$$

\* Rank = no. of pivot columns ( $r$ )

Nullity = no. of free columns ( $n-r$ )  
dimension of null space

$n = \text{total no. of columns in augmented coefficient matrix}$

$$\boxed{\text{Rank} + \text{Nullity} = \text{Total no. of columns}}$$

word meaning of word

System of linear equations

Homogeneous

$$Ax=0$$

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

consistent eqns

no solution  
inconsistent eqns

Find echelon form  
of augmented matrix

if [0 0 0 | non-zero] exists

if [0 0 0 | non-zero] doesn't exist

free variables?

Yes  
infinite solutions

No  
unique soln

$$\left[ \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \xrightarrow{\text{Echelon form}} \left[ \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \xrightarrow{\text{Augmented matrix}}$$

(non homogeneous)  
Heterogeneous

$$Ax=b$$

No. of linearly independent solutions = no of free variables

$\text{rank}(A) = \text{rank}(A|b)$   $\Rightarrow$  Nullity

$$\left[ \begin{array}{c|c} A & b \end{array} \right]$$

$$\text{Rank}(A) = \text{Rank}(A|b) \Leftrightarrow$$

$b$  is Linear combination  
of columns of  $A$ .

∴ solution exists.

$\text{rank}(A|b) = \text{rank}(A)$   $\Rightarrow$  rank  $A$  is full rank

$$\text{Rank}(A) \neq \text{Rank}(A|b) \Leftrightarrow$$

$$\text{Rank}(A|b) = \text{Rank}(A) + 1$$

$b$  is not linear combination  
of columns of  $A$ .

∴ solution does not exist

Find  $\text{Rank}(A)$  &  
 $\text{Rank}(A|b)$

$$\text{Rank}(A) \neq \text{Rank}(A|b)$$

$\text{rank}(A|b) = \text{rank}(A)$   $\Rightarrow$  Rank(A) =

$$\text{Rank}(A|b)$$

Inconsistent  
equations  
no solution

zero or more zero rows

zero or more zero columns

Consistent eqns.

$$\text{Rank}(A) = n ?$$

$$\left[ \begin{array}{c|ccccc} 1 & * & * & * & * & * \\ 0 & * & * & * & * & * \\ 0 & * & * & * & * & * \\ 0 & * & * & * & * & * \end{array} \right] \quad \text{Yes}$$

unique  
solution

No

Infinite  
solution

Given:-  $A \rightarrow mxn$  - matrix

①  $\text{rank}(A) = m = n$

$$\left[ \begin{array}{cccc|c} 1 & x & x & x & \\ 0 & 1 & x & x & \\ 0 & 0 & 1 & x & \\ 0 & 0 & 0 & 1 & \end{array} \right] \quad \left[ \begin{array}{c|c} b & \\ \hline A & b \end{array} \right]$$

$\Leftrightarrow$  (A|b) has a unique solution  
rank(A) = rank(A|b) = n  
every column has pivot  
 $\therefore \text{rank}(A) = \text{rank}(A|b) = n$

$\Leftrightarrow$  (A|b) has a unique solution  
unique solution

rank(A) = rank(A|b)  $\Leftrightarrow$  (A|b) has a unique solution

$\therefore$  A has unique solution

A has few free variables

②  $\text{rank}(A) = m$

$m \neq n$

$$\left[ \begin{array}{cccc|c} 1 & x & x & x & x & \\ 0 & 1 & x & x & x & \\ 0 & 0 & 1 & x & x & \\ 0 & 0 & 0 & 1 & x & \end{array} \right] \quad \left[ \begin{array}{c|c} b & \\ \hline (A|b) & b \end{array} \right]$$

$\Leftrightarrow$  (A|b) has a unique solution  
 $\therefore$  There exist free variables  
infinite solutions

$\Leftrightarrow$  (A|b) has a unique solution

$\because m \neq n \therefore$  can't be unique solution  
free variable will always be present

③  $\text{rank}(A) = n$

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

$\Leftrightarrow$  0 or 1 solutions

$\text{rank}(A) = \text{rank}(A|b)$  if  $0000|0$  exists  
unique solution (no free variables)

$\text{rank}(A) \neq \text{rank}(A|b)$  if  $0000|0$  non zero exists  
no solution

⑧ ⑨  $\text{rank}(A) < m$  &  $\text{rank}(A) < n$   $\Rightarrow$  no soln

$$\left[ \begin{array}{c|c} \begin{matrix} 0 \\ 0 \\ 0 \end{matrix} & \begin{matrix} 0 \\ 0 \\ 0 \end{matrix} \end{array} \right] \quad \text{if } 0 \text{ then, infinite soln}$$

( $\because$  free variable is present)

i.e. if not zero, no solution.

and when  $m = n$  or  $\infty$  solutions determined by

$$\text{rank}(A) = \text{rank}(B) \Rightarrow \text{rank}(A) = \text{rank}(B) = r$$

\* If all columns are linearly independent, then only trivial solution exists for  $AX = 0$ .

$A_{m \times n}$ ,  $\text{rank}(A) = n$

$$\left[ \begin{array}{c|c} \begin{matrix} w & b \\ w+d & b+r \end{matrix} & \begin{matrix} x \\ y \end{matrix} \end{array} \right] \xrightarrow{AX=0} \left[ \begin{array}{c|c} \begin{matrix} w & b \\ w+d & b+r \end{matrix} & \begin{matrix} 0 \\ 0 \end{matrix} \end{array} \right]$$

From previous

$$\left[ \begin{array}{c|c} \begin{matrix} w & b \\ w+d & b+r \end{matrix} & \begin{matrix} x \\ y \end{matrix} \end{array} \right] \xrightarrow{\text{add } d(A)} \left[ \begin{array}{c|c} \begin{matrix} w & b \\ w+d & b+r \end{matrix} & \begin{matrix} x+d \\ y+r \end{matrix} \end{array} \right]$$

Now add  $d(A)$  to both sides

$$\left[ \begin{array}{c|c} \begin{matrix} w & b \\ w+d & b+r \end{matrix} & \begin{matrix} x+d \\ y+r \end{matrix} \end{array} \right] \xrightarrow{\text{add } d(A)} \left[ \begin{array}{c|c} \begin{matrix} dd & dr \\ dd & dr \end{matrix} & \begin{matrix} dd & dr \\ dd & dr \end{matrix} \end{array} \right]$$

## Chapter 2: Determinants, Eigenvalues & eigenvectors

<all matrices are square matrices>

$\text{Det}(A) \leftarrow \text{scalar}$

### Properties -

①  $\text{Det}(\text{Identity matrix}) = 1$

② Determinant changes sign when 2 rows are exchanged.

1 mark  
question in  
gate 2023

$$\det \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = -1 \quad \det \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = 1$$

$$③ \begin{vmatrix} a+p & b+q \\ c+r & d+w \end{vmatrix} = \begin{vmatrix} a & b \\ c & d \end{vmatrix} + \begin{vmatrix} p & q \\ r & w \end{vmatrix}$$

Linearity of  
one row at a  
time

$$\begin{vmatrix} a+p & b+q \\ c+r & d+w \end{vmatrix} = \begin{vmatrix} a & b \\ c+r & d+w \end{vmatrix} + \begin{vmatrix} p & q \\ c+r & d+w \end{vmatrix}$$

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} + \begin{vmatrix} a & b \\ r & w \end{vmatrix} + \begin{vmatrix} p & q \\ c & d \end{vmatrix} + \begin{vmatrix} p & q \\ r & w \end{vmatrix}$$

$$\rightarrow \begin{vmatrix} ta & tb \\ tc & td \end{vmatrix} = t \begin{vmatrix} a & b \\ tc & td \end{vmatrix} = t^2 \begin{vmatrix} a & b \\ c & d \end{vmatrix}$$

$$2 \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} 2a & 2b \\ 2c & 2d \end{bmatrix}$$

$$2 \begin{vmatrix} a & b \\ c & d \end{vmatrix} = 2 \begin{bmatrix} 2a & 2b \\ c & d \end{bmatrix} = \begin{vmatrix} 2a & b \\ 2c & d \end{vmatrix}$$

①  $|A+B| \neq |A| + |B|$

② Determinant of diagonal matrix is product of diagonal elements.

③ Number of terms in the determinant of  $n \times n$  matrix

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

For  $2 \times 2$  mat.  $\rightarrow 2! = 2$  terms ( $a-d, b-c$ )  
 For  $3 \times 3$  mat.  $\rightarrow 3! = 6$  terms.

### Imp. properties of determinants -

④  $\det(AB) = \det(A) * \det(B)$

⑤  $\det(A^{-1}) = \frac{1}{\det(A)}$

⑥  $\det(A^T) = \det(A)$

Sign of cofactor of  $a_{ij} = (-1)^{i+j}$

without sign  $\rightarrow$  minor  $\rightarrow \begin{vmatrix} a & b \\ c & d \end{vmatrix}$

with sign  $\rightarrow$  cofactor  $\rightarrow (-1)^{i+j} \begin{vmatrix} a & b \\ c & d \end{vmatrix}$

\* If elements of a row / column are multiplied with cofactors of any other row / column, then, their sum is 0.

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

$$A^{-1} = \frac{1}{|A|} \begin{bmatrix} C_{11} & C_{12} & C_{13} \\ C_{21} & C_{22} & C_{23} \\ C_{31} & C_{32} & C_{33} \end{bmatrix}^T$$

(coefficient matrix)<sup>T</sup>

adjoint of  $A = \begin{bmatrix} C_{11} & C_{12} & C_{13} \\ C_{21} & C_{22} & C_{23} \\ C_{31} & C_{32} & C_{33} \end{bmatrix}^T$

$$A^{-1} = \frac{1}{|A|} \text{ adj } A$$

$$\frac{1}{(A)_{lab}} = ({}^T A)_{lab}$$

$$(A)_{lab} = ({}^T A)_{lab}$$

(-1)  $\in P$  matrix property

$|A| \neq 0$  for non-singular

## Crammer's rule

→ mathematics of millionaires.

• very expensive method of finding solutions of  $AX=b$ .

more resources / computing power required.

→ only for theory purpose.

$$AX = b \Rightarrow X = A^{-1} \cdot b = \frac{1}{|A|} \begin{bmatrix} C_1 \\ C_2 \\ C_3 \end{bmatrix} = \frac{1}{|A|} \begin{bmatrix} C_{11} & C_{21} & C_{31} \\ C_{12} & C_{22} & C_{32} \\ C_{13} & C_{23} & C_{33} \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

$$= \frac{1}{|A|} \begin{bmatrix} b_1 C_{11} + b_2 C_{21} + b_3 C_{31} \\ b_1 C_{12} + b_2 C_{22} + b_3 C_{32} \\ b_1 C_{13} + b_2 C_{23} + b_3 C_{33} \end{bmatrix} = \frac{1}{|A|} \det \left( \begin{bmatrix} b_1 & a_{12} & a_{13} \\ b_2 & a_{22} & a_{23} \\ b_3 & a_{32} & a_{33} \end{bmatrix} \right)$$

$$= \frac{1}{|A|} \det \left( \begin{bmatrix} a_{11} & b_1 & a_{13} \\ a_{21} & b_2 & a_{23} \\ a_{31} & b_3 & a_{33} \end{bmatrix} \right)$$

$$= \frac{1}{|A|} \det \left( \begin{bmatrix} a_{11} & a_{12} & b_1 \\ a_{21} & a_{22} & b_2 \\ a_{31} & a_{32} & b_3 \end{bmatrix} \right)$$

~~STRAIGHT DT WHICH !!~~

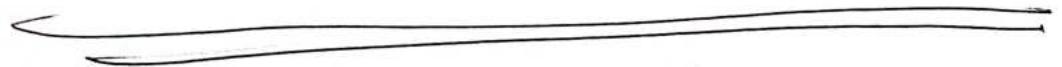
$$X_1 = \frac{1}{\det \left( \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \right)}$$

Very expensive.

$$\det \left( \begin{bmatrix} b_1 & a_{12} & a_{13} \\ b_2 & b_{22} & b_{23} \\ b_3 & b_{32} & b_{33} \end{bmatrix} \right)$$

$$\det \left( \begin{bmatrix} a_{11} & b_1 & a_{13} \\ a_{21} & b_2 & a_{23} \\ a_{31} & b_3 & a_{33} \end{bmatrix} \right)$$

$$\det \left( \begin{bmatrix} a_{11} & a_{12} & b_1 \\ a_{21} & a_{22} & b_2 \\ a_{31} & a_{32} & b_3 \end{bmatrix} \right)$$



Imp ques

Give conditions when

$$2x_1 + hx_2 = k$$

$$x_1 - x_2 = 2$$

have (i) Infinite soln

(ii) No soln (iii) Unique soln

$$\begin{bmatrix} 2 & h \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} k \\ 2 \end{bmatrix}$$

i) For infinite solutions —

↳ b should be LC of columns of A

↳ columns of A should be L.D.

$$h = -2 \text{ and } k = 4$$

ii) For no solution —

↳ columns of A should be L.D.

↳ b should not be LC of columns of A

$$\therefore h = -2 \text{ and } k \neq 4$$

iii) for unique solutions

↳ b should be L.C. of columns of A

RREF and cond.

A satisfy

at RREF

2 LI vectors in  $\mathbb{R}^2$

can generate anything

COLLINESITY

for 2 vectors in  $\mathbb{R}^2$  —

→ linearly dependent

→ unique soln

→ no soln

→ infinite soln

→ linearly independent

→ unique soln

→ no soln

→ infinite soln

→ linearly independent

→ unique soln

→ no soln

→ linearly independent

→ unique soln

→ no soln

→ linearly independent

→ unique soln

→ no soln

→ linearly independent

→ unique soln

→ no soln

→ linearly independent

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→ no soln

→ linearly independent

→ unique soln

→ no soln

→ linearly independent

→ unique soln

# Eigenvalues and eigenvectors

(only square matrix)

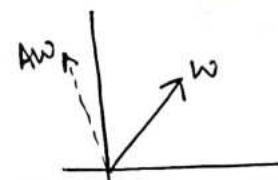
$$Ax = \lambda x$$

eigen vector of  $A$

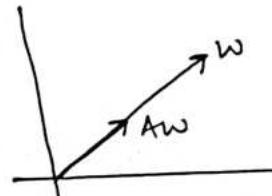
vector gets scaled.

scalar

$\lambda = 1$	no change
$\lambda > 1$	stretched
$\lambda < 1$	shrunken



vector  $w$  is not eigen vector of vector  $A$



vector  $w$  is eigen vector of  $A \cdot X$

$\lambda = 0.5$ .  
eigen value

several properties of matrices can be analysed based on their eigen values.

$|A - \lambda I| = 0$  (roots of characteristic)

## Application

core concept of Regularization is eigen value & eigen vectors

Machine learning cannot survive without this

## Principal Component Analysis

By definition, eigen vectors are non-zero.

$$Av = \lambda v$$

matrix

eigen vector

eigen value

eigen values ~~can~~ ~~not~~ be can be used

for other uses as well

types II & III are used

- exists non-zero bns and solving identically
- There are infinitely many eigen vectors for every matrix  $A$ .  
(vector space part)

No. of LI eigen vectors =  $n$  for In.

Linearly independent

For  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ , no. of LI eigen vectors = 3.

- If  $x$  is eigenvector, then, any non zero multiple of  $x$  is also an eigenvector.

spans all zero multiple of  $1 > k$   
dimensions  $1 > k$

Characteristic equation :-  $|A - \lambda I| = 0$

Eigen vectors from different eigen values are linearly independent.

$\lambda_1$  &  $\lambda_2$  are eigenvalues.

If  $\lambda_1 \neq \lambda_2 \rightarrow$  corresponding eigen vectors are linearly independent.

If  $\lambda_1 = \lambda_2 \rightarrow$  if  $\lambda_1$  is repeating 2 times, then, we can have e.g. either one or 2 LI eigen vectors.

No. of LI eigen vectors for following characteristic equation -

$$(\lambda-1)^2 (\lambda-5) (\lambda-7)^3 = 0$$

- (a) atleast 6      (c) exactly 6  
~~(b) almost 6~~      ~~(d) atleast 3~~  
(e) exactly 3.

$$(\lambda-x_1)^{m_1} (\lambda-x_2)^{m_2} (\lambda-x_3)^{m_3} \dots \dots (\lambda-x_k)^{m_k}$$

↳ atleast k LI eigen vectors

↳ almost  $m_1 + m_2 + m_3 + \dots + m_k = n$  all LI eigen vectors

↳ geometric multiplicity  $\leq$  arithmetic multiplicity  $\Rightarrow$  almost m LI eigen vectors

↳ geometric multiplicity  $\leq$  algebraic multiplicity  $\Rightarrow$  almost m LI eigen vectors

① Geometric multiplicity:  
no. of LI eigenvectors corresponding to  $\lambda = \lambda_1$ .

$$A \cdot M \cdot (I - \lambda I)^{-1} \geq (G \cdot M \cdot (I - \lambda I))^{-1}$$

Rank of matrix II  $\leq$  rank of product matrix

Rank of matrix II  $\leq$  rank of product matrix

Rank of matrix II  $\leq$  rank of product matrix

Condition when  $AM = \underline{AM}$ :

i.e. if  $(\lambda - \lambda_1)^m$ , then, what should be the value of  $\lambda_1$  such that there are  $m$  LI eigenvectors for  $\lambda = \lambda_1$ .

↳ picture (2)

↳ picture (2)

Geometric multiplicity: Number of linearly independent eigenvectors corresponding to  $\lambda$ .

Arithmet.

Algebraic multiplicity: Number of times  $\lambda$  is repeating

$$\text{AM} \leq A \cdot M$$

water says II & fractal ↗

→ Real symmetric matrices have  $\text{G.M.} = \text{A.M.}$

↳ n real eigen values.

↳ n orthogonal eigenvectors.

this concept  
is needed  
in diagonalization  
in machine  
learning!!

In real symmetric matrices, all eigen vectors are LI even if eigen values are repeating.

For real symmetric matrix

$$(\lambda-1)^3 (\lambda-5)^7 (\lambda-11) = 0$$

11 LI eigen vectors

Corresponding to  $\lambda=1$ , 3 LI eigen vectors

" "  $\lambda=5$ , 7 LI " "

" "  $\lambda=11$ , 1 LI " "

~~Imp~~ Two important properties of eigen values

ये properties मूलन की इजाजत नहीं है

समझे?

2023 में दुम

मूलन की खता कर चुके हो!!



① Determinant of a matrix is product of eigen values.

② Trace (sum of elements of main diagonal) of matrix is sum of eigen values.

$$|A| = \lambda_1 \lambda_2 \lambda_3 \dots \lambda_n \quad \text{or} \quad |A| = (\lambda_1 \lambda_2 \dots \lambda_n)$$

प्रत्येक गणितीय विश्लेषण में trace(A) =  $\lambda_1 + \lambda_2 + \lambda_3 + \dots + \lambda_n$

③ If 2 rows of A are same, one of the eigen values is 0.

$$\begin{bmatrix} 3 & 1 & 4 \\ 1 & 1 & 5 \\ 1 & 1 & 5 \end{bmatrix}$$

R<sub>2</sub> & R<sub>3</sub> are same

∴ one of the eigenvalues is 0.

Q = A is true

$$0 = XA \quad \leftarrow \quad Q = X[I - A]$$

If columns of a matrix are linearly dependent  $\Rightarrow$  determinant of the matrix is 0.

$\Rightarrow$  product of eigen values is 0.

$\Rightarrow$  one of the eigen values is 0.

$$0 = XA$$

$$0 = X(I - A)$$

Q = XA

$$[J_3 + J_2] = 0$$

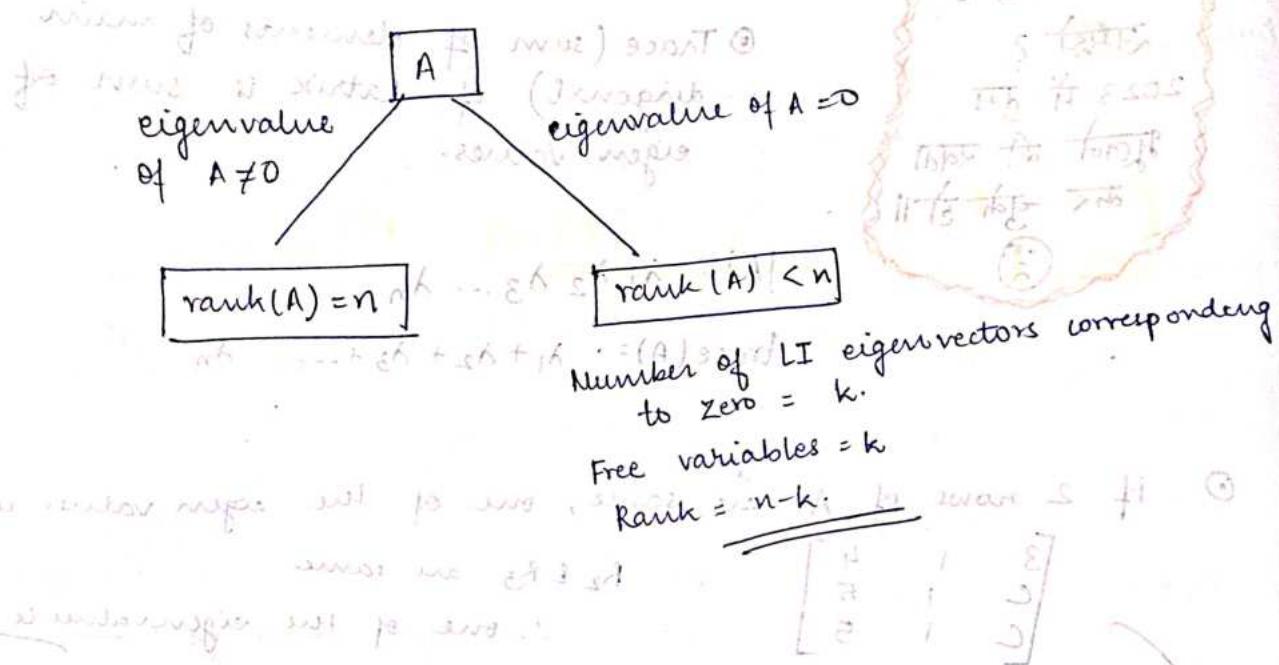
विकास सर्वे द्वारा

$$8 = 8 - 8 = 0 \rightarrow \text{det}(A) = 0$$

S = 0

## Relationship between rank and eigenvalues

If one of the eigen values of  $A \neq 0 \Rightarrow \text{rank}(A) = n$   
 If one of the eigen values of  $A = 0 \Rightarrow \text{rank}(A) < n$



when  $\lambda=0$ ,

$$[A - \lambda I]x = 0 \rightarrow Ax = 0$$

i.e. it is similar to solving  $Ax = 0$ . Similar to

$\therefore \text{rank} = n - \text{no. of LI vectors}$

$\therefore$   $\text{rank} = n - 2$  (for 2 free variables)

$\therefore$   $\text{rank} = n - 2$  (for 2 free variables)

No. of LI eigenvectors for  $\lambda=0$  are 2. What is the rank of matrix.

$$(A - \lambda I)x = 0 \quad \because \lambda=0 \quad \therefore Ax = 0$$

$A_{10 \times 10}$

solution has 2 LI eigenvectors

multiplicity = 2

$$\text{rank} = n - 2 = 10 - 2 = 8$$

i.e. it is of the form

$$s[\ ] + t[\ ]$$

re. 2 free variables

$(A - \lambda I)x = 0$  and  $Ax = 0$  can be connected by  $\lambda = 0$  ↑

$\Rightarrow A^T = A$  (symmetric matrix)

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

## Cayley-Hamilton Theorem

Every square matrix satisfies its own characteristic equation.

### Application

- To calculate positive integral powers of A.
- To calculate inverse of a square matrix.

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} = V \quad \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} = W = I$$

Imp. If  $\lambda$  is the eigenvalue of  $AB$ , then, it is also the eigenvalue of  $BA$ .

$$AB \cdot x = \lambda x$$

$$\Rightarrow B \cdot ABx = B\lambda x \Rightarrow (BA)(Bx) = \lambda(Bx)$$

i.e.  $AB$  and  $BA$

share non zero eigen values.

make just sure  $Bx \neq 0$

$Bx$  is the eigenvector of  $BA$ .  
 $\lambda$  is eigen value of  $BA$ .

inferred.

~~A~~  
 $A_{2 \times 10}$

~~B~~  
 $B_{10 \times 2}$

~~C~~  
 $C_{2 \times 4}$

~~D~~  
 $D_{4 \times 1}$

$AB_{2 \times 2} \rightarrow$  eigen values = 1, 5

$BA_{10 \times 10} \rightarrow$  eigen values = 1, 5, 0, 0, 0, 0, 0, 0, 0, 0  
 coz only non-zero eigen values are shared.  
 remaining should be 0.

~~A~~  
 $A_{4 \times 3}$

$B_{3 \times 4}$

$AB_{4 \times 4}$  cannot have 4 non zero eigen values.  
 one of the eigen values should be 0.  
 coz.  $BA_{3 \times 3}$  can have only 3 eigen values & all non zero e.v. should be shared.  
 ∴ not possible if 4 non zero e.v. are there for  $AB$ .

Imp question

If  $u = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \end{bmatrix}_{6 \times 1}$  &  $v = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \\ 2 \\ 6 \end{bmatrix}_{6 \times 1}$

also  $\lambda$  largest eigen value of  $uv^T$   
 non zero eigen values of  $uv^T =$  non zero eigen values of  $v^Tu$

$(uv)^T u = (v^T u)u$   $\Rightarrow v^T u = 26.8A$

As  $u$  is orthogonal to  $v$   
  $u^T v = 0$   
  $u^T u = 1^2 + 2^2 + 3^2 + 4^2 + 5^2 + 6^2 = 91$

$\therefore \text{Ans} = 17$

## Eigen values of powers of A

- ① If  $\lambda_1, \lambda_2$  are eigen values of  $A$ , then eigen values of  $A^k$  are  $\lambda_1^k, \lambda_2^k$ .

- ② eigen value of  $f(A)$  is  $f(\lambda)$

if  $\lambda$  is eigen value of  $A$ , then eigen value of  $A^k + 3A^{k-1} + I$  is  $\lambda^k + 3\lambda^{k-1} + 1$

Ques

Suppose  $x$  is an eigenvector of  $A$  such that  $Ax = \lambda_1 x$  and  $x$  is also eigenvector of  $B$  such that  $Bx = \lambda_2 x$ . What is the eigen value of  $(A + \frac{1}{2}B)^{-1}$ ?

- A)  $\frac{\lambda_1}{2\lambda_1 + \lambda_2}$       B)  $\frac{\lambda_2}{2\lambda_1 + \lambda_2}$       C)  $\frac{2}{2\lambda_1 + \lambda_2}$       D)  $\frac{\lambda_1}{2\lambda_2 + \lambda_1}$

$$(A + \frac{1}{2}B)x = Ax + \frac{1}{2}Bx = \lambda_1 x + \frac{1}{2}\lambda_2 x$$

$$\text{eigen values for } A + \frac{1}{2}B = \left(\lambda_1 + \frac{\lambda_2}{2}\right)x$$

$\therefore$  eigen value of  $A + \frac{1}{2}B$  is  $\lambda_1 + \frac{\lambda_2}{2}$

$\therefore$  eigen value of  $(A + \frac{1}{2}B)^{-1}$  is  $\left[\begin{array}{c} \lambda_1 + \frac{\lambda_2}{2} \\ \hline \end{array}\right]^{-1}$

$$\begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 5 \\ 1 & 5 & 8 \end{bmatrix} = 0$$

$$\begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 5 \\ 1 & 5 & 8 \end{bmatrix} = 0$$

$$\begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 5 \\ 1 & 5 & 8 \end{bmatrix} = 0$$

- ③ If eigen values are imaginary — then, its conjugate will also be eigen value.  
i.e. if  $2+i$  is e.v. of  $A$  then,  $2-i$  is also eigen value.

## LU Decomposition

Decomposing a matrix into 2 parts - L and U  
 where L = lower triangular matrix with 1 at diagonal

$U$  = upper triangular matrix.

$$\begin{bmatrix} 1 & 4 & -3 \\ -2 & 8 & 5 \\ 3 & 4 & 7 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ * & 1 & 0 \\ * & * & 1 \end{bmatrix} \begin{bmatrix} x & x & x \\ 0 & x & x \\ 0 & 0 & x \end{bmatrix} U.$$

Method: None of the relationships are in the form  $w = kx^m$

⑧ Convex + A) into snow echelon form with SW

swapping the rows is (d) not (A)

④ The  $A$  is now echelon form gives upper

$$\text{triangular matrix} \cdot x = x \left( \frac{1}{\lambda} A + I \right)$$

⑥ The corresponding negative of coefficients applied to rows in operations gives lower triangular matrix.

$$A = \begin{bmatrix} 1 & 4 & -3 \\ -2 & \left( \frac{-2}{-2} + 1 \right) & \frac{5}{-2} \\ 3 & -4 & 7 \end{bmatrix}$$

$$R_2 \rightarrow R_2 + 2R_1$$

$$\begin{bmatrix} 1 & 4 & -3 \\ 0 & 16 & -1 \\ 3 & 4 & 7 \end{bmatrix}$$

$$R_3 \rightarrow R_3 - 3R_1$$

$$\begin{bmatrix} 1 & 4 & -3 \\ 0 & 16 & -1 \\ 0 & -8 & 16 \end{bmatrix}$$

$$R_{2y} \rightarrow R_{2y} + \frac{1}{2}R_2$$

$$\left[ \begin{array}{ccc} 1 & 4 & -3 \\ 0 & 16 & -1 \\ 0 & 0 & 15.5 \end{array} \right]$$

$$L_2 = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 3 & -\frac{1}{2} & 1 \end{bmatrix}$$

11

## Types of matrices

1. Identity matrix — Is at principal diagonal 1's everywhere else.

2. Inverse of a matrix —  $A$  and  $B$  are inverse of each other if

$$AB = BA = I_n$$

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

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

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

3. Transpose of a matrix — Transpose of a matrix is an operator that flips a matrix over its diagonal.

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}^T = \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix}$$

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

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

$$3. (AB)^T = B^T A^T$$

$$4. (cA)^T = cA^T$$

5. dot product of 2 column vectors can be calculated by

$$a \cdot b = a^T b$$

$$6. (A^T)^{-1} = (A^{-1})^T$$

associates w.r.t

#### 4. Triangular matrix.

① lower triangular if all the elements above its main diagonal are zero.  $a_{ij} = 0 \quad i > j$

② upper triangular if all the elements below its main diagonal are zero.  $a_{ij} = 0 \quad i < j$

$$AT = A^T = AA$$

→ Product of 2 lower triangular matrices is lower triangular matrix  $AB = BA$

→ Product of 2 upper triangular matrices is upper triangular matrix  $A^T B^T = (BA)^T$

#### 5. Diagonal matrix

all the entries outside the main diagonals are zero. matrix is of diagonal form

③ if  $i \neq j$ ,  $a_{ij} = 0$

$$\begin{bmatrix} s & & \\ & s & \\ & & s \end{bmatrix} = A^T \begin{bmatrix} s & & \\ & s & \\ & & s \end{bmatrix}$$

#### 6. Symmetric matrix

matrix equal to its own transpose

$$A = A^T \quad A^T = T(TA) = A$$

$$T_B + T_A = T(B + A) = A + B$$

#### 7. Skew symmetric matrix

elements of principal diagonal are zero

$$\begin{bmatrix} 0 & 2 & -45 \\ -2 & 0 & -4 \\ 45 & 4 & 0 \end{bmatrix} = T(A)$$

① For any matrix A, both  $A^T$  and  $A^TA$  are symmetric.

② For any matrix A with real no. entries,

$A + A^T$  is symmetric

$A - A^T$  is skew symmetric

$$A = \frac{1}{2} \left[ (A + A^T) + (A - A^T) \right]$$

sym      skew

symmetric

skew symmetric

③ Every matrix can be represented as a sum of symmetric & skew symmetric matrix.

Symmetrized form of  $A$  is  $\frac{1}{2}(A + A^T)$  and skew form of  $A$  is  $\frac{1}{2}(A - A^T)$ .

$$\frac{1}{2}(A + A^T) + \frac{1}{2}(A - A^T)$$

Ques:- If A and B are symmetric, then,  $ABA$  is

A. Symmetric

B. Skew symmetric

C. Diagonal

D. Triangular.

(Given:-  $A^T = A$ ,  $B^T = B$ )

$$(ABA)^T = A^T B^T A^T$$

$$= B A^T A$$

∴ Symmetric

Ans: D

Ans: D

## 8. Orthogonal matrix

Orthogonal matrix is a real square matrix whose columns and rows are orthogonal vectors

$$Q^T Q = I$$

If a matrix is orthogonal, then, its

transpose is equal to its inverse.

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

$A_{nn}$  matrix is diagonalizable only if it has  $n$  LI eigen vectors

$$\overline{[(TA - A) = (TA + A)]} \Rightarrow A^2 = A$$

Ques

Consider the matrix  $A_{nxn}$  having the following characteristic equation

$$\lambda^2(\lambda-3)(\lambda+2)^3(\lambda-4)^3$$

as  $A^2A$  exist, differentiable end & true  $\lambda$  if 2nd

What could be rank of  $A$ ?

A. 6

$$TA - TAT^{-1} = T(A^2A)$$

C. 8

D. 9

differentiable  $\Rightarrow$  rank  $\Rightarrow$  corresponding to  $\lambda=0$ ,  
No. 6 decompose eigen vector  
Rank 1 2 LI eigen vector  
Rank 2 3 LI eigen vector

thus in the case

Nullity = 1 or 2

$\therefore$  Rank = 9-1 or 9-2  
8 or 7

rank decompose 0

rank decompose 0 is rank decompose  
rank decompose 0 rank decompose 0

Solution is not unique when  $A$  is not unit  $\Rightarrow$  column of  $A$  are L.D.

rank decompose 0 rank decompose 0

The eigen vectors corresponding to eigen values of real symmetric matrix are orthogonal.

$$\sum_{i=1}^n \sum_{j=1}^n |A_{ij}|^2 = \text{trace}(A \bar{A})$$

# PROBABILITY

- conditional probability
- Probability distribution
  - random variable
  - Expectation

## Inclusion Exclusion Principle

For any 2 events E and F,

$$P(E \cup F) = P(E) + P(F) - P(E \cap F)$$

For any 3 events E, F & G,

$$P(E \cup F \cup G) = P(E) + P(F) + P(G) - P(E \cap F) - P(F \cap G) - P(E \cap G) + P(E \cap F \cap G)$$

## (De Morgan's Rule for Probability)

$$P((E \cap F)^c) = P(E^c \cup F^c)$$

$$P((E \cup F)^c) = P(E^c \cap F^c)$$

## Conditional Probability

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$P(A \cap B) = P(B) P(A|B) = P(A) P(B|A)$$

$$P(A \cap B \cap C) = P(A, B, C) = P(C) P(B|C) P(A|B, C) = P(A) P(B|A) P(C|A, B)$$

$$= P(B) P(A|B) P(C|A, B)$$

$$= P(B) P(C|B) P(A|C, B)$$

two events A  $\leftarrow$  A  $\sqcup$  A

shows independence

$$(A \cap B \cap C)^c + (A \cap B \cap C) = (A \cap C)^c$$

$$\Sigma P(A_i) x_i + \Sigma P(A_i) x_i =$$

$$x = (A)^c$$

$$(A \cap B \cap C)^c = (A \cap B \cap C)^c$$

$$= (A \cap B \cap C)^c$$

## Marginalization (also called Total Probability)

Let  $A_1, A_2$  and  $A_3$  partition the sample space then,

for any event  $B$ ,

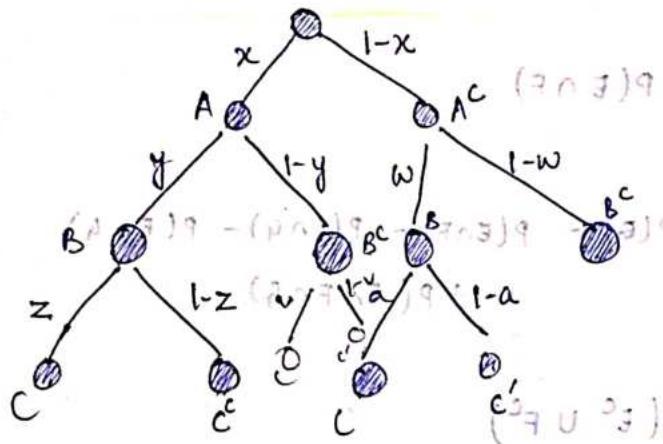
$$P(B) = P(B|A_1) + P(B|A_2) + P(B|A_3)$$

$$P(B) = P(B) \cap P(B|A_1) + P(B|A_2) + P(B|A_3)$$

$$P(B) = P(B) \cap P(B|A_1) + P(B|A_2) + P(B|A_3)$$

$$P(B) = P(B) \cap P(B|A_1) + P(B|A_2) + P(B|A_3)$$

$$P(B) = P(B) \cap P(B|A_1) + P(B|A_2) + P(B|A_3)$$



$$P(A \cap B \cap C) = xyz$$

$$P(A \cup B \cup C) = x + y + z - (x \cdot y) - (y \cdot z) - (x \cdot z) + (x \cdot y \cdot z)$$

$$P(C|A, B) = z$$

$$P(C|B) = \frac{P(B \cap C)}{P(B)}$$

$$P(B \cap C) = P(A \cap B \cap C) + P(A^c \cap B \cap C)$$

$$= (x \cdot y \cdot z) + ((1-x) \cdot w \cdot z)$$

$$= (1-x)(w+z)$$

$$P(B) = P(A \cap B) + P(A^c \cap B)$$

$$= (x \cdot y) + (1-x) \cdot w$$

$$P(C|A) = \frac{P(C \cap A)}{P(A)}$$

$$P(C \cap A) = P(A \cap B \cap C) + P(A \cap B^c \cap C)$$

$$= xyz + x(1-y)z$$

$$P(A) = x$$

$$\therefore P(C|A) = \frac{xyz + x(1-y)z}{x} = yz + (1-y)z$$

**Independent Events**

$$P(B|A) = P(B)$$

$$P(A \cap B) = P(A) P(B)$$

$$P(A \cap B \cap C) = P(A) P(B) P(C)$$

$A \perp\!\!\!\perp B \rightarrow A$  and  $B$  are independent events.

3 events  $A, B$  &  $C$  are independent if all below

following conditions are satisfied -

$$P(A \cap B) = P(A) P(B)$$

$$P(B \cap C) = P(B) P(C)$$

$$P(A \cap C) = P(A) P(C)$$

$$P(A \cap B \cap C) = P(A) P(B) P(C)$$

## Conditional independence

conditional making

2 events A & B are independent given C -

$$P(A \cap B | C) = P(A|C) \cdot P(B|C)$$

and same for B & C

$$\{ \text{or} \geq (0) \} = P(A|B,C) \cdot P(B|C) \geq x$$

$$= P(B|A,C) \cdot P(A|C)$$

→ ~~independent condition~~ ~~independent condition~~ are ~~independent~~ part ③

\* It is possible that earlier A & B were independent

but after C, they got dependent

e.g. ~~P(A ∩ B) = P(A) P(B) = x~~ ~~∴ P(A ∩ B | C) = P(A|C) P(B|C)~~

Independence does not imply conditional independence  
and vice versa.

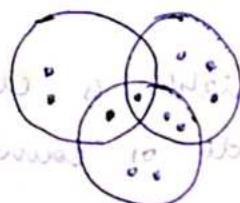
conditional making for part

$$P(A \cap B | C) = P(A|C) \cdot P(B|C)$$

but

$$P(A \cap B) \neq P(A) P(B)$$

∴ below



Two events A and B are independent -

$$P(A \cap B) = P(A) P(B)$$

$$P(A|B) = P(A)$$

$$P(B|A) = P(B)$$

If A & B are independent,

A & B<sup>c</sup> are independent

A<sup>c</sup> & B are independent

A<sup>c</sup> & B<sup>c</sup> are independent.

Two events A & B are conditionally independent -

$$P(A \cap B | C) = P(A|C) P(B|C)$$

$$P(A|B,C) = P(A|C)$$

$$P(B|A,C) = P(B|C)$$

If A & B are conditionally independent given C, then,

A & B<sup>c</sup> are independent given C

A<sup>c</sup> & B are independent given C

A<sup>c</sup> & B<sup>c</sup> are independent given C

## Random variables

Some properties of random variables

Function that maps an outcome to the real number  
 $(\Omega, \mathcal{A}) \ni (\omega, A) \mapsto (\Omega, \mathcal{A}, \mathbb{P})$

$x < 10 \in \{\omega | X(\omega) < 10\} \rightarrow$  set of some elements  
= event.

$(\Omega, \mathcal{A}) \ni (\omega, A) \mapsto$

① Any condition on random variables is an event.

Events are sets of outcomes that satisfy a condition.

$\rightarrow X$  maps any outcome to either  $a, b$ , or  $c$ .

$(\Omega, \mathcal{A}) \ni (\omega, A) \mapsto$  Then,  $x = a, x = b$ , and  $x = c$ . There are 3 events are

mutually exclusive

Some properties of random variables

Types of Random variables

$(\Omega, \mathcal{A}) \ni (\omega, A) \mapsto (\Omega, \mathcal{A}, \mathbb{P})$

1. Discrete - Random variable is called discrete if it takes either finite or countably infinite no. of values.

Integers

Ex → no. of sixes in 2 rolls.

Ans 0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, ...

Properties of random variables

2. Continuous - Random variable takes an uncountably infinite no. of values.

$(\Omega, \mathcal{A}) \ni (\omega, A) \mapsto (\Omega, \mathcal{A}, \mathbb{P})$

Ex → choosing a point from  $[0, 1]$

Properties of random variables are

• well defined

• well behaved

• well distributed

• well behaved

Properties of random variables

Properties of random variables

Properties of random variables

Properties of random variables

## Probability Mass Function

Listing down probability of each value for a discrete random variable is called 'Probability Mass Function'.

## Expectation

→ single number that summarises PMF.

→ weighted average (in proportion to probabilities) of the possible values of  $X$ .

$$E(X) = \sum_k k P(X=k)$$

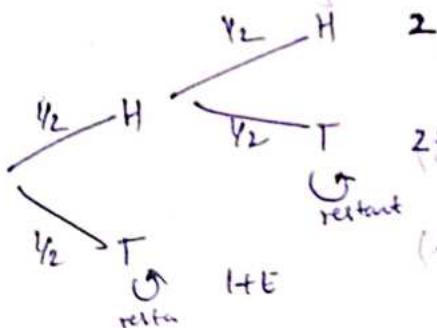
Expectation is just average on enough data.

As amount of data increases, Average  $\rightarrow$  Expectation.

$$\textcircled{Q} \rightarrow (3+1) \frac{1}{5} + (13+1) \frac{1}{5} = 3$$

(Gate 2015)

Let the random variable  $X$  represent the no of times a fair coin needs to be tossed till 2 consecutive heads appear for the first time. The expectation of  $X$



$$\begin{aligned}
 & \text{Expectation} = \\
 & E = 2 \times \frac{1}{4} + \frac{1}{4} \times (2+E) + \frac{1}{2} \times (1+E) \\
 & \Rightarrow E = \frac{1}{2} + \frac{2+E+2+E}{4} = \frac{1}{2} + \frac{4+3E}{4} \\
 & \Rightarrow \frac{4E-4-3E}{4} = \frac{1}{2} \Rightarrow \frac{E-4}{4} = \frac{1}{2} \\
 & \Rightarrow E-4=2 \Rightarrow \underline{\underline{E=6}}
 \end{aligned}$$

~~Gate 2005~~

rechts mit ziemlich gut

An unbiased coin is tossed repeatedly until the outcome of 2 successive tosses is same. Assuming here that each toss is independent, the expected number of trials are

number of tosses is

- A. 3  
B. 4  
C. 5  
D. 6

$$\begin{array}{c} \text{H} \xrightarrow{\frac{1}{2}} \text{H} \quad \text{I} \\ \text{T} \xrightarrow{\frac{1}{2}} \text{T} \quad (1+E_2) \end{array}$$

$\text{H}$  means no change in sequence  
 $\text{T}$  means we have to start again

$$\therefore E_1 = \frac{1}{2} + \frac{1}{2}(1+E_2) \quad \textcircled{1}$$

MT 2020-21

of no. of tosses to get HH or TT

$\times$  to consider

$\frac{1}{2} \xrightarrow{\text{H}} \text{T} \quad (1+E_2)$

$(\text{H}=\text{X}) \Rightarrow \frac{1}{2} \xrightarrow{\text{X}} \text{H}$  expected

let  $E_1 = \text{no. of H tosses to get HH when we have H in hand}$

$E_2 = \text{no. of expected tosses to get TT when we have T in hand}$

$$E = \frac{1}{2}(1+E_1) + \frac{1}{2}(1+E_2) \quad \textcircled{3}$$

~~Now~~  $\text{H} \xrightarrow{\frac{1}{2}} \text{H} \quad (1+E_1)$

from  $\textcircled{1}$  &  $\textcircled{2}$ ,

$$E_1 = \frac{1}{2} + \frac{1}{2}(1+E_1) \quad \textcircled{2}$$

$$E_1 + E_2 = 1 + \frac{1}{2}(2 + E_1 + E_2) = 2 + \frac{E_1 + E_2}{2}$$

$\Rightarrow E_1 + E_2 = 2$  with trial no. of

$\frac{2}{2} = 1$

Now,

$$E = \frac{1}{2}(1+E_1) + \frac{1}{2}(1+E_2)$$

$$(2+1) \times \frac{1}{2} + (2+2) \times \frac{1}{2} + \frac{1}{2} = \frac{1}{2} + \frac{1}{2} + \frac{1}{2}(E_1 + E_2)$$

$$\frac{3S+3}{2} + \frac{1}{2} = \frac{3S+3+3+2}{2} + \frac{1}{2} \underbrace{E_1 + E_2}_{2} = 1 + 2 = 3$$

$$\frac{1}{2} = \frac{1-3}{2} \quad (\because \frac{1}{2} = \frac{3S-2-3N}{2})$$

$$1-3 = 3-2 \quad (\because S=N-3)$$

## Variance

- ① Variance is average of squared distance from mean.
- ② always non negative

③ If variance is low, then, for any random data point we can expect value of that datapoint to be very close to what we expect!

$$\cancel{[d]_{\text{cov}} + [x_0]_{\text{cov}}} = [d]^2 + [x_0]^2 = [d+x_0]^2 = [y]^2$$

	Calculate using data	Calculate using probability distribution
Expectation	$\bar{x} = \frac{x_1 + x_2 + x_3 + \dots + x_n}{n}$	$= E[X] = \sum k P(X=k) = \sum k P(X=k)$
Variance	squared distance from mean $\sigma^2 = \frac{(x_i - \bar{x})^2}{n}$	$E((X - E[X])^2)$
	$\sigma^2 = \text{Var}(X)$	(Q2) machine learning?

$$\text{Var}(X) = E[(X - E[X])^2]$$

$$\text{Var}(X) = E[X^2] - (E[X])^2$$

$$\text{Var}[x_1 + 2x_2 + 3x_3 + 4x_4]$$

$$= \text{Var}[x_1] + \text{Var}[2x_2] + \text{Var}[3x_3] + \text{Var}[4x_4]$$

$$= \text{Var}[x_1] + 2^2 \text{Var}[x_2] + 3^2 \text{Var}[x_3] + 4^2 \text{Var}[x_4]$$

only valid if  $x_1, x_2, x_3, x_4$  are independent

## Properties

### Properties of expectations and variance

Let  $Y = ax + b$

$$\textcircled{1} \quad E[Y] = E[ax+b] = E[ax] + E[b] = aE[x] + \cancel{E[b]}$$

~~Properties of expectation~~  
Properties of variance

$$\textcircled{2} \quad \text{Var}[Y] = \text{Var}[ax+b] = \text{Var}[ax] + \cancel{\text{Var}[b]}^0$$

$$= a^2 \text{Var}[x]$$

$$\textcircled{3} \quad E[x_1 + x_2] = E[x_1] + E[x_2]$$

$$\text{Var}[x_1 + x_2] = \text{Var}[x_1] + \text{Var}[x_2] = \bar{x}$$

[if  $x_1$  &  $x_2$  are independent]

$$(E[X] - \bar{x})$$

$$\text{Var}[x_1 - x_2] = \text{Var}[x_1] + \text{Var}[x_2]$$

Standard deviation (SD) =  $\sqrt{\text{Variance}}$

$$\text{Covariance } \text{Cov}(X, Y) = E[(\bar{x} - E[X])(\bar{y} - E[Y])]$$

$$\text{Cov}(X, Y)^2 \leq \frac{(E[X]^2) - E[X]^2}{\text{Var}(X) \text{Var}(Y)} = (\bar{x}) \text{cov}$$

$$[EXN(\text{cov}) + E[XE] \text{cov} + E[X^2] \text{cov} + E[X]^2 \text{cov}] \text{cov}$$

$$[EX(\text{cov})^2 + E[X](\text{cov})^2 + E[X^2] \text{cov} - E[X]^2 \text{cov} + E[X]^2 \text{cov}] =$$

# Discrete Random variable

- Bernoulli (only 2 possible outcomes — 0 or 1)
- Binomial
- Poisson
- Uniform

$$\frac{f(x)}{P(X=x)} = (1-p)^q$$

(Success x)

## 1. Bernoulli Random variable -

If we have some experiment [where we can classify every outcome as success or failure] then, we can say that we have bernoulli distribution

To Be Remembered

$$\begin{cases} E[X] = 1 \cdot p + 0 \cdot (1-p) = p & P(X=1) = p \\ E[X^2] = 1^2 \cdot p + 0^2 \cdot (1-p) = p & \\ \text{var}[X] = E[X^2] - (E[X])^2 = p - p^2 = p(1-p) & \end{cases}$$

PMF =  $X = \begin{cases} 1 & \text{with probability } p \\ 0 & \text{with probability } 1-p \end{cases}$

## 2. Binomial Random Variable

Repeated independent trials of Bernoulli

Performing experiment  $n$  times

No. of success =  $k$ ; No. of failures =  $n-k$

$P(X=k) = {}^n C_k \cdot p^k \cdot (1-p)^{n-k}$

$E[X] = np$

$\text{var}[X] = np(1-p) = npq$

PG 35 - 10.57 etc.

3. Poisson distribution

( $n \rightarrow$  very large)  $p \rightarrow$  very small  
 parameter  $\lambda$  binomial distribution becomes Poisson distribution

$$P(X=k) = \frac{e^{-\lambda} \lambda^k}{k!} \quad k=0, 1, 2, \dots, \infty$$

[ $k$  successes]

- additive property followed  
 effects are as follows  $E[X] = \sum_{k=0}^{\infty} k \cdot P(X=k) = \sum_{k=0}^{\infty} k \cdot \frac{e^{-\lambda} \lambda^k}{k!} = \lambda$   
 ex. first accident to happen is in uniform phase

No individual measurement event in day (pos. no.)

$$\text{var}[X] = E[X^2] - (E[X])^2 = \lambda$$

$$i = (i \neq X)^q$$

$$q = (q-1)q + q^2 = [X]_3$$

$$q = (q-1) \cdot q + q^2 = [X]_3$$

$$(q-1)q = q^2 - q = [X]_3 - [X]_2 = [X]_{10}$$

Ques: If on an average there are 2 accidents per day.

Ques: What is the probability of 4 accidents on given day?

$$\lambda = 2$$

$$P(X=4) = \frac{e^{-\lambda} \lambda^4}{4!} = \frac{e^{-2} 2^4}{4!} = \frac{16}{24} e^{-2} = \frac{2}{3 e^2}$$

Illustrated for white bacteriophages infection

Based on successive binomial

Some patterns where poisson distribution is to be used

→ No. of misprints on a page

→ No. of people in a community surviving to 100.

→ No. of wrong telephone nos dialed in a day.

→ No. of ...

$$pq^N + (q-1)q^N = [X]_{100}$$

$$\text{Rate} \quad \frac{pq^N}{N} = \frac{pq^N}{N(N-1)} = \frac{pq^N}{N^2 - N}$$

#### Uniform Discrete Random variable

A random variable  $X$  has a discrete uniform distribution if each of the  $n$  values in its range, say,  $x_1, x_2, \dots, x_n$  has equal probability.

$$\text{i.e. } P(x_i) = \frac{1}{n}$$

where  $P(x)$  represents the probability mass function (PMF)

~~Due~~  
If  $X$  is uniformly distributed over  $\{a, a+1, \dots, b\}$   
then find out

$$P(X=x) = \frac{1}{b-a+1}$$

$$E(x)$$

$$\text{Var}(x)$$

No. of elements,  $n = b-a+1$

$$B: P(X=x) = \frac{1}{b-a+1}$$

$$E(x) = \frac{1}{b-a+1} [a + a+1 + \dots + b-1 + b]$$

$$= \frac{1}{b-a+1} \left[ \frac{b(b+1)}{2} - \frac{a(a-1)}{2} \right] = \frac{1}{b-a+1} \left[ \frac{b^2 - a^2 + b + a}{2} \right]$$

$$= \frac{1}{b-a+1} \left[ \frac{(b-a)(b+a) + (b+a)}{2} \right] = \frac{\frac{b-a+1}{b-a+1} \cdot \frac{b+a}{2}}{\frac{b-a+1}{b-a+1}} = \frac{b+a}{2}$$

variance does not change on shifting R.V. to left or right by a constant

$$\therefore \text{Var}(x+a) = \text{Var}(x)$$

required variance of  $x = \text{variance of } \{1, 2, \dots, b-a+1\}$

$$E[x^2] = \frac{1}{n} (1^2 + 2^2 + \dots + n^2) = \frac{n(n+1)(2n+1)}{6 \cdot n} = \frac{(n+1)(2n+1)}{6}$$

$$E[x] = \frac{1}{n} (1+2+3+\dots+n) = \frac{1}{n} \cdot \frac{n(n+1)}{2} = \frac{n+1}{2} = \frac{(b-a+1)^2 - 1}{12}$$

## Continuous Random variable

Continuous random variable  $X$  follows number A  
 or is said to have distribution if its probability density function is

- Uniform
- Exponential
- Normal

### Probability Density Function

Random variable  $X$  has a PDF  $f(x)$  if ~~P(X=x)~~

where probability is  $P(a \leq X \leq b) = \int_a^b f(x) dx$  for all  $a, b$ .

(TMQ)  $\int_a^a f(x) dx = 0$

For valid PDF,

$$P(X \leq \infty) = \int_{-\infty}^{\infty} f(x) dx = 1$$

$$P(X \leq b) = \int_{-\infty}^b f(x) dx$$

### 1. Uniform Continuous Random variable

Probability  $\propto$  length of interval

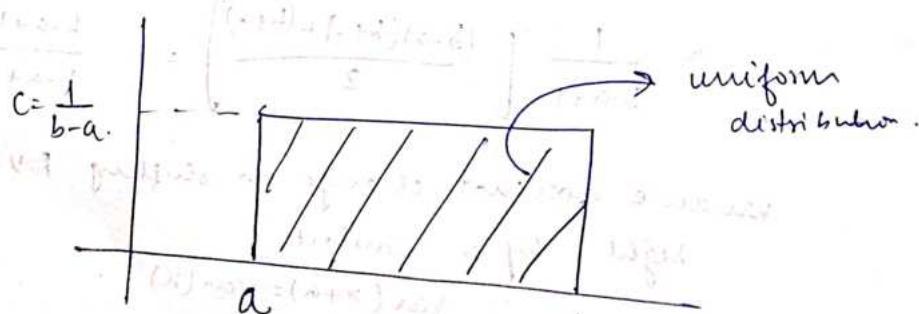
$$f(x) = \begin{cases} c & \text{if } a \leq x \leq b \\ 0 & \text{otherwise} \end{cases}$$

Prob.

$$P(\text{interval}) = c(\text{length of interval})$$

$$c = \frac{1}{b-a}$$

$$E(X) = \frac{a+b}{2}$$



$$\text{Var}(X) = \frac{(b-a)^2}{12}$$

$$\int_3^6 f(x) dx = \int_6^7 f(x) dx \quad \text{cuz interval is raw.}$$

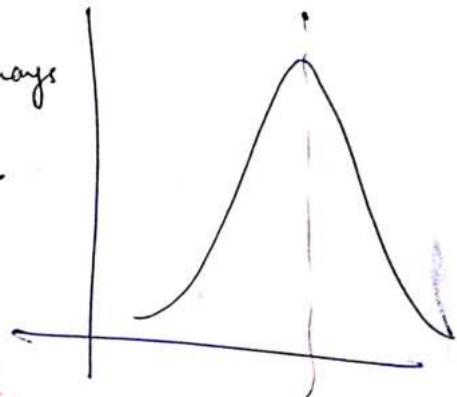
## 2. Normal Distribution / Gaussian Distribution

symmetrical distribution (about mean)

$$E[x] = \mu$$

$$\text{Var}[x] = \sigma^2$$

$\mu$  &  $\sigma^2$  are always given in question



Standard normal distribution  $\rightarrow \sigma=1, \mu=0$

The given distribution  $(X)$  is converted into standard normal distribution  $(Z)$

$$* Z = \frac{x-\mu}{\sigma}$$

$$E[X] = \mu$$

$$\text{Var}[X] = \sigma^2$$

$$E[Z] = 0$$

$$\text{Var}[Z] = 1$$

## 3. Exponential distribution

continuous random variable whose probability density function is given by

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & x < 0 \end{cases}$$

$$E[X] = \frac{1}{\lambda}$$

$$\text{Var}[X] = \frac{1}{\lambda^2}$$

Calculus

$\lim_{x \rightarrow a} f(x) = f(a)$  if  $f(a)$  does not exist, it exists only if

$$\lim_{n \rightarrow a} f(x) = \lim_{n \rightarrow a} f(n)$$

$$\text{LHL} = \lim_{x \rightarrow a^-} f(x) = \lim_{h \rightarrow 0} f(a-h)$$

$$\text{RHL} = \lim_{x \rightarrow a^+} f(x) = \lim_{h \rightarrow 0} f(a+h)$$

let  $\lim_{x \rightarrow a} f(x) = l$  and  $\lim_{n \rightarrow a} g(n) = m$

$$1. \lim_{x \rightarrow a} (f(x) \pm g(x)) = \lim_{x \rightarrow a} f(x) \pm \lim_{x \rightarrow a} g(x) = l \pm m$$

$$2. \lim_{x \rightarrow a} (f(x) \times g(x)) = \lim_{x \rightarrow a} f(x) \times \lim_{x \rightarrow a} g(x) = l \times m$$

$$3. \lim_{x \rightarrow a} \frac{f(x)}{g(x)} = \frac{\lim_{x \rightarrow a} f(x)}{\lim_{x \rightarrow a} g(x)} = \frac{l}{m}$$

$$4. \lim_{x \rightarrow a} (f(x))^n = \lim_{x \rightarrow a} f(x)^{n-a} \lim_{x \rightarrow a} g(x) = l^n \text{ if } \lim_{x \rightarrow a} f(x) = l > 0$$

$$5. \lim_{x \rightarrow a} f(g(x)) = f(\lim_{x \rightarrow a} g(x)) = f(m) \text{ provided } f(x) \text{ is continuous at } g(x)=m$$

### Frequently used limits -

$$\lim_{x \rightarrow 0} \frac{\sin x}{x} = 1 \quad \lim_{x \rightarrow 0} \frac{e^x - 1}{x} = 1 \quad \lim_{x \rightarrow 0} \frac{\log(1+x)}{x} = 1$$

$$* \lim_{x \rightarrow 0} \frac{a^x - 1}{x} = \log_a e$$

1<sup>o</sup> form

$$1 \cdot l = \lim_{x \rightarrow a} (1 + f(x))^{g(x)}$$

$$\lim_{x \rightarrow a} g(x) = \infty$$

$$\lim_{x \rightarrow a} f(x) = 0$$

1<sup>o</sup> form

$$\therefore \log l = \lim_{x \rightarrow a} g(x) \log (1 + f(x))$$

$$= \lim_{x \rightarrow a} g(x) \frac{\log (1 + f(x))}{f(x)} \cdot f(x)$$

$$= \lim_{x \rightarrow a} g(x) \cdot \lim_{x \rightarrow a} \frac{\log (1 + f(x))}{f(x)} \cdot \lim_{x \rightarrow a} f(x)$$

$$= \lim_{x \rightarrow a} g(x) f(x)$$

$$\therefore l = e^{\lim_{x \rightarrow a} g(x) f(x)}$$

$$2 \cdot l = \lim_{x \rightarrow a} (f(x))^g(x) \quad \lim_{x \rightarrow a} f(x) = 1 \quad \lim_{x \rightarrow a} g(x) = \infty \quad [1^\infty \text{ form}]$$

$$\log l = \lim_{x \rightarrow a} g(x) \log f(x) = \lim_{x \rightarrow a} g(x) \log (1 + f(x) - 1)$$

$$= \lim_{x \rightarrow a} g(x) \cdot \frac{\log (1 + f(x) - 1)}{f(x) - 1} \cdot (f(x) - 1)$$

$$= \lim_{x \rightarrow a} g(x) \frac{\log (1 + f(x) - 1)}{f(x) - 1} \quad \lim_{x \rightarrow a} f(x) = 1$$

$$= \lim_{x \rightarrow a} g(x) \cdot (f(x) - 1)$$

$$\therefore l = e^{\lim_{x \rightarrow a} g(x) (f(x) - 1)}$$

## Imp

① If  $f(x)$  and  $g(x)$  are continuous at  $x=a$ , then,

①  $f(x) \pm g(x)$  is also continuous at  $x=a$

②  $f(x) \cdot g(x)$  &  $\frac{f(x)}{g(x)}$  are also continuous at  $x=a$ .

$f(g(x))$  is continuous at  $x=a$ .

If  $g(x)$  is continuous at  $x=a$ ,

then  $f(x)$  is continuous at  $g(a)$ .

## Differentiability

A function  $f(x)$  is differentiable at  $x=a$ , iff

left hand derivative = Right hand derivative

$$\text{LHD} = \text{RHD}$$

$$\text{i.e. } \lim_{h \rightarrow 0} \frac{f(a+h) - f(a)}{h} = \lim_{h \rightarrow 0} \frac{f(a) - f(a-h)}{h}$$

## Basic properties and formulae -

let  $f(x)$  &  $g(x)$  be differentiable functions  $c$  &  $n$  real nos.

$$① (cf)' = cf'(x)$$

$$⑤ \frac{d}{dx} c = 0$$

$$② (f \pm g)' = f'(x) \pm g'(x)$$

$$⑥ \frac{d}{dx} x^n = nx^{n-1}$$

$$③ (fg)' = f'g + g'f$$

$$⑦ \frac{d}{dx} f(g(x)) = f'(g(x)) g'(x)$$

$$④ \left(\frac{f}{g}\right)' = \frac{f'g + fg'}{g^2}$$

## Common derivatives

- ①  $\frac{d}{dx}(x) = 1$       ⑥  $\frac{d}{dx}(a^x) = a^x \ln a$
- ②  $\frac{d}{dx}(\sin x) = \cos x$       ⑦  $\frac{d}{dx}(e^x) = e^x$
- ③  $\frac{d}{dx}(\cos x) = -\sin x$       ⑧  $\frac{d}{dx}(\ln(x)) = \frac{1}{x}$
- ④  $\frac{d}{dx}(\tan x) = \sec^2 x$       ⑨  $\frac{d}{dx} \log a^x = \frac{1}{x \ln a}$
- ⑤  $\frac{d}{dx}(\sec x) = \sec x \tan x$

## Intermediate value Theorem

If  $f$  is continuous function in the closed interval  $[a, b]$ , and if  $d$  is b/w  $f(a)$  and  $f(b)$ , then there is a number  $c \in [a, b]$  with  $f(c) = d$ .

## Rolle's Theorem

If a function  $f$  is

continuous in  $[a, b]$

differentiable in  $(a, b)$

$f(a) = f(b)$

then, there exists  $c$  in  $(a, b)$  such that  $f'(c) = 0$

## Ineterminate forms in limit

$\frac{\infty}{\infty}$	$\frac{0}{0}$	$\infty - \infty$	$0^0$	$0 \cdot \infty$	$\infty^0$	$1^\infty$
①	②	③	④	⑤	⑥	⑦

$\curvearrowleft$   
L'Hopital rule

$\downarrow$   
taking log

$L = \lim_{x \rightarrow a} f(x)^{g(x)}$  form  
 $L = e^{\lim_{x \rightarrow a} g(x)(f(x)-1)}$   
 (Taking log)

# Integration

- ①  $\int x^n dx = \frac{x^{n+1}}{n+1} + C \quad n \neq -1$
- ②  $\int \frac{1}{x} dx = \ln|x| + C$
- ③  $\int e^x dx = e^x + C$
- ④  $\int a^x dx = \frac{a^x}{\ln a} + C$
- ⑤  $\int \sin x dx = -\cos x + C$
- ⑥  $\int \cos x dx = \sin x + C$

## Properties of definite integral

$$\int_a^b f(x) dx = \int_a^b f(t) dt$$

$$\int_a^b f(x) dx = - \int_b^a f(x) dx$$

$$\int_a^b f(x) dx = \int_a^c f(x) dx + \int_c^b f(x) dx$$

$$\int_a^b f(x) dx = \int_a^b f(a+b-x) dx$$

$$\int_{-a}^a f(x) dx = 0 \quad \text{if } f(x) \text{ is odd function i.e. } f(-x) = -f(x)$$

$$\int_{-a}^a f(x) dx = 2 \int_0^a f(x) dx \quad \text{if } f(x) \text{ is even function i.e. } f(-x) = f(x)$$

$$\int u v dx = u \int v dx - \int \left( \frac{du}{dx} \int v dx \right) dx$$

however good for integration by parts