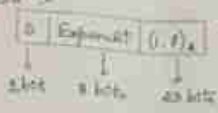


FLOATING POINT REPRESENTATION AND ERROR

bits
 $(\text{exponent} - \text{bias}) \times (2^{\pm \text{signification}})$
 $(\text{exponent}) \times (2 - \text{sign})$
 of Mantissa =



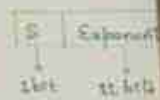
Convert given number to binary number

If given number is
 $+^n$, then $S=0$.
 $-^n$, then $S=1$.

Convert value of C to binary number



bits
 $\times 2^{\text{exponent}} \times (1 - \text{sign})$
 of Mantissa =



0 or 1
 2^{exponent}

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

Find $f'(x)$
 Find $f(x_0), f(x_1), f(x_2), f(x_3)$
 If consecutive $+^n$ and $-^n$ root lies between, accept a $-^n$ value of $f(x)$.
 For substitute value to the
 $x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$
 Continue this process until we get the number.

Method

Bisection Method

$x_{n+1} = x_n - \left[\frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} \right] f(x_n)$
 $f(x_0), f(x_1), f(x_2), \dots$ with
 give $+^n$ and $-^n$ numbers
 between accept $+^n$ and
 number as x_n and
 the first approximation
 its value to the form
 $x_n = \left[\frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} \right] f(x_n)$
 $f(x)$ For x
 process until we get

Secant

GAUSS ELIMINATION METHOD

Consider a system of linear equations
 $a_{11}x + a_{12}y + \dots + a_{1n}z = b_1$
 $a_{21}x + a_{22}y + \dots + a_{2n}z = b_2$
 \dots
 $a_{m1}x + a_{m2}y + \dots + a_{mn}z = b_m$

Equivalent to matrix equation
 $Ax = b$
 $\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} x \\ y \\ \dots \\ z \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \dots \\ b_m \end{bmatrix}$

Thinking as per matrix equation
 $A^{-1}Ax = A^{-1}b$
 $x = A^{-1}b$

PARTIAL AND COMPLETE PIVOTING

A row pivot element with row and column pivot pivoting
 Selection on basis last column and row for the largest element
 Column complete pivoting

CHOLESKY METHOD

It is used to solve linear system of equations
 $Ax = b$
 where $A = [a_{ij}]$ is symmetric and positive definite
 $A = LL^T$
 where $L = [l_{ij}]$ is lower triangular
 and L^T is upper triangular
 and L is unit lower triangular
 and L^T is unit upper triangular

CONVERGENCE ANALYSIS OF ITERATIVE METHOD

A general linear iteration method for the solution of the system of equations
 $Ax = b$
 is given by
 $x^{(k+1)} = Bx^{(k)} + c$
 where $B = I - \alpha A$
 and $c = \alpha b$
 The iteration converges if
 $\rho(B) < 1$
 where $\rho(B)$ is the spectral radius of B
 For $A = [a_{ij}]$ and $B = [b_{ij}]$
 $b_{ij} = \delta_{ij} - \frac{a_{ij}}{a_{ii}}$
 and $c_i = \frac{b_i}{a_{ii}}$
 For $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ and $B = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$
 $\rho(B) = 0 < 1$
 So the iteration converges

SYSTEM OF LINEAR EQUATIONS

INTRODUCTION
 Observations from Algebraic system
 Common in various fields of science and engineering to solve some set of linear equations by iterative methods and matrix method.

ROWER METHOD

This method is used to compute the inverse of a matrix and also the solution of the system of linear equations
 $Ax = b$
 where $A = [a_{ij}]$ is a square matrix
 and $b = [b_i]$ is a column vector
 The inverse of A is given by
 $A^{-1} = [a^{-1}_{ij}]$
 and the solution is given by
 $x = A^{-1}b$

ITERATIVE METHODS

JACOBI'S METHOD

GAUSS-SEIDEL METHOD

NUMERICAL SOLUTION OF ORDINARY DIFFERENTIAL EQUATIONS

Consider a system of first order ordinary differential equations
 $\frac{dx}{dt} = f(t, x)$
 with initial condition
 $x(0) = x_0$

TRIANGULARIZATION METHOD

This method is also known as decomposition method or the LU decomposition method
 Consider a square matrix $A = [a_{ij}]$
 $A = LU$
 where L is lower triangular
 and U is upper triangular
 The above system can be solved by
 $LUx = b$
 $Ux = L^{-1}b$
 where L^{-1} is lower triangular
 and $L^{-1}b$ is a column vector
 The system can be solved by
 $Ux = c$
 where $c = L^{-1}b$
 and U is upper triangular
 The system can be solved by
 $x = U^{-1}c$
 where U^{-1} is upper triangular
 and $U^{-1}c$ is a column vector

It is used to solve numerical problems
 reduced solving requirements
 Solution for matrix A and vector b
 the matrix is lower triangular
 and the vector is known
 the solution is given by
 $x = U^{-1}c$
 where U^{-1} is upper triangular
 and $U^{-1}c$ is a column vector

SYSTEM OF LINEAR EQUATIONS

METHOD 1: Gaussian Elimination

The system of linear equations is written as $Ax = b$ where A is the coefficient matrix, x is the vector of unknowns, and b is the constant vector.

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

Step 1: Write the augmented matrix $[A|b]$ as $[C|d]$ where $C = [a_{ij}]$ and $d = [b_i]$.

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

Step 2: Use row operations to transform the augmented matrix into row echelon form (REF) or reduced row echelon form (RREF).

$$\left[\begin{array}{cccc|c} 1 & 0 & \dots & 0 & r_1 \\ 0 & 1 & \dots & 0 & r_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & r_n \\ 0 & 0 & \dots & 0 & 0 \end{array} \right]$$

Method 2: Gauss-Jordan

Step 1: Use row operations to transform the augmented matrix into reduced row echelon form (RREF).

$$\left[\begin{array}{cccc|c} 1 & 0 & \dots & 0 & r_1 \\ 0 & 1 & \dots & 0 & r_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & r_n \\ 0 & 0 & \dots & 0 & 0 \end{array} \right]$$

Step 2: The system of linear equations is now in RREF.

Step 3: Use the back substitution method to solve for the unknowns.

Step 4: The solution is given by $x = [x_1, x_2, \dots, x_n]^T$.

Method 3: Inverse Matrix

Method 1: Inverse Matrix Method. The system of linear equations $Ax = b$ can be solved by multiplying both sides by the inverse of A , A^{-1} .

Method 2: Inverse Matrix Method. The system of linear equations $Ax = b$ can be solved by multiplying both sides by the inverse of A , A^{-1} .

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

Method 3: Inverse Matrix Method. The system of linear equations $Ax = b$ can be solved by multiplying both sides by the inverse of A , A^{-1} .

$$x = \frac{1}{\det(A)} \begin{bmatrix} \dots \\ \dots \\ \dots \end{bmatrix}$$

Method 4: Partial Pivoting

Method 1: Partial Pivoting. The system of linear equations $Ax = b$ can be solved by using partial pivoting to avoid large round-off errors.

Method 2: Partial Pivoting. The system of linear equations $Ax = b$ can be solved by using partial pivoting to avoid large round-off errors.

Method 3: Partial Pivoting. The system of linear equations $Ax = b$ can be solved by using partial pivoting to avoid large round-off errors.

Method 4: Partial Pivoting. The system of linear equations $Ax = b$ can be solved by using partial pivoting to avoid large round-off errors.

Method 5: LU Decomposition

Method 1: LU Decomposition. The system of linear equations $Ax = b$ can be solved by decomposing A into L and U .

Method 2: LU Decomposition. The system of linear equations $Ax = b$ can be solved by decomposing A into L and U .

$$Ax = b \Rightarrow LUx = b \Rightarrow Ux = L^{-1}b$$

Method 3: LU Decomposition. The system of linear equations $Ax = b$ can be solved by decomposing A into L and U .

$$x = U^{-1}L^{-1}b$$

Method 6: Power Method

Method 1: Power Method. The dominant eigenvalue of a matrix A can be found by repeatedly multiplying a vector x by A .

Method 2: Power Method. The dominant eigenvalue of a matrix A can be found by repeatedly multiplying a vector x by A .

Method 3: Power Method. The dominant eigenvalue of a matrix A can be found by repeatedly multiplying a vector x by A .

Method 4: Power Method. The dominant eigenvalue of a matrix A can be found by repeatedly multiplying a vector x by A .

$$x_{k+1} = \frac{Ax_k}{\|Ax_k\|}$$

Method 5: Power Method. The dominant eigenvalue of a matrix A can be found by repeatedly multiplying a vector x by A .

Method 7: QR Method

Method 1: QR Method. The eigenvalues of a matrix A can be found by decomposing A into Q and R .

Method 2: QR Method. The eigenvalues of a matrix A can be found by decomposing A into Q and R .

$$A = QR$$

Method 3: QR Method. The eigenvalues of a matrix A can be found by decomposing A into Q and R .

$$x = R^{-1}Q^{-1}Ax$$

INTERPRETING AND NUMERICAL DISCRIMINATION

Example 1
 20 - 2 = 18
 100 - 20 = 80
 100 - 20 = 80

Example 2
 100 - 20 = 80
 100 - 20 = 80

Example 3
 100 - 20 = 80
 100 - 20 = 80

Example 4
 100 - 20 = 80
 100 - 20 = 80

Example 5
 100 - 20 = 80
 100 - 20 = 80

Example 6

Example 7

Example 8

Example 9

100 - 20 = 80
 100 - 20 = 80
 100 - 20 = 80

100 - 20 = 80
 100 - 20 = 80
 100 - 20 = 80

100 - 20 = 80
 100 - 20 = 80
 100 - 20 = 80

100 - 20 = 80
 100 - 20 = 80
 100 - 20 = 80

Decimal	Binary	Octal	Hexadecimal
0	0000	0	0
1	0001	1	1
2	0010	2	2
3	0011	3	3
4	0100	4	4
5	0101	5	5
6	0110	6	6
7	0111	7	7
8	1000	10	8
9	1001	11	9
10	1010	12	A
11	1011	13	B
12	1100	14	C
13	1101	15	D
14	1110		E
15	1111		F

Binary Representation
0, 1, 2, 3, 4, 5, 6, 7, 8

Decimal Representation
0, 1, 2, 3, 4, 5, 6, 7, 8, 9, A, B, C, D, E, F

Decimal Representation
Decimal form of base 10 representation.
On 0.1, 0.2, ... 0.9, 1.0, 2.0, ...
The notation 1.234 and its decimal equivalent.

Floating Point Representation

0.00000001	1.00000000
0.00000010	0.00000001
0.00000011	0.00000010
0.00000100	0.00000100
0.00000101	0.00000101
0.00000110	0.00000110
0.00000111	0.00000111
0.00001000	0.00001000
0.00001001	0.00001001
0.00001010	0.00001010
0.00001011	0.00001011
0.00001100	0.00001100
0.00001101	0.00001101
0.00001110	0.00001110
0.00001111	0.00001111

FLOATING POINT REPRESENTATION AND ERRORS

ERROR
Non-zero additive change due to operations due to finite number.

Machine Representation
The way an internal computer uses floating point numbers to represent a number.

1. Rounding Errors
Arithmetic involving a number which with infinite precision can be approximated by a floating point number.

2. Mathematical Goals
The user of the floating point number is interested in the result of the operation, not the internal representation of the number.

Types of Errors
1. Rounding Error
2. Truncation Error
3. Catastrophic Cancellation

Absolute Error
The error in a computed quantity of desired accuracy is the absolute error.
Approximate value - Exact value = Absolute Error

Relative Error
The relative error of a number is the absolute error divided by the true value.
Error / True Value = Relative Error

3. Results of Errors
The error in a computed quantity of desired accuracy is the absolute error.
Approximate value - Exact value = Absolute Error

4. Machine Errors
The error in a computed quantity of desired accuracy is the absolute error.
Approximate value - Exact value = Absolute Error

5. Truncation Error
The error in a computed quantity of desired accuracy is the absolute error.
Approximate value - Exact value = Absolute Error

Machine and Mathematical Goals
The user of the floating point number is interested in the result of the operation, not the internal representation of the number.

1. Using Taylor series expansion
$$1 + x \approx 1 + x - \frac{x^2}{2} + \frac{x^3}{6} - \dots$$

2. Using Taylor series expansion
$$1 + x \approx 1 + x - \frac{x^2}{2} + \frac{x^3}{6} - \dots$$

3. Using Taylor series expansion
$$1 + x \approx 1 + x - \frac{x^2}{2} + \frac{x^3}{6} - \dots$$

4. Using Taylor series expansion
$$1 + x \approx 1 + x - \frac{x^2}{2} + \frac{x^3}{6} - \dots$$

5. Using Taylor series expansion
$$1 + x \approx 1 + x - \frac{x^2}{2} + \frac{x^3}{6} - \dots$$

6. Using Taylor series expansion
$$1 + x \approx 1 + x - \frac{x^2}{2} + \frac{x^3}{6} - \dots$$

7. Using Taylor series expansion
$$1 + x \approx 1 + x - \frac{x^2}{2} + \frac{x^3}{6} - \dots$$

2. Machine Representation
The floating point number is represented by a sign, a mantissa, and an exponent.
Example: 1.234×10^5

3. Machine Representation
The floating point number is represented by a sign, a mantissa, and an exponent.
Example: 1.234×10^5

4. Machine Representation
The floating point number is represented by a sign, a mantissa, and an exponent.
Example: 1.234×10^5

5. Machine Representation
The floating point number is represented by a sign, a mantissa, and an exponent.
Example: 1.234×10^5

6. Machine Representation
The floating point number is represented by a sign, a mantissa, and an exponent.
Example: 1.234×10^5

7. Machine Representation
The floating point number is represented by a sign, a mantissa, and an exponent.
Example: 1.234×10^5

8. Machine Representation
The floating point number is represented by a sign, a mantissa, and an exponent.
Example: 1.234×10^5

9. Machine Representation
The floating point number is represented by a sign, a mantissa, and an exponent.
Example: 1.234×10^5

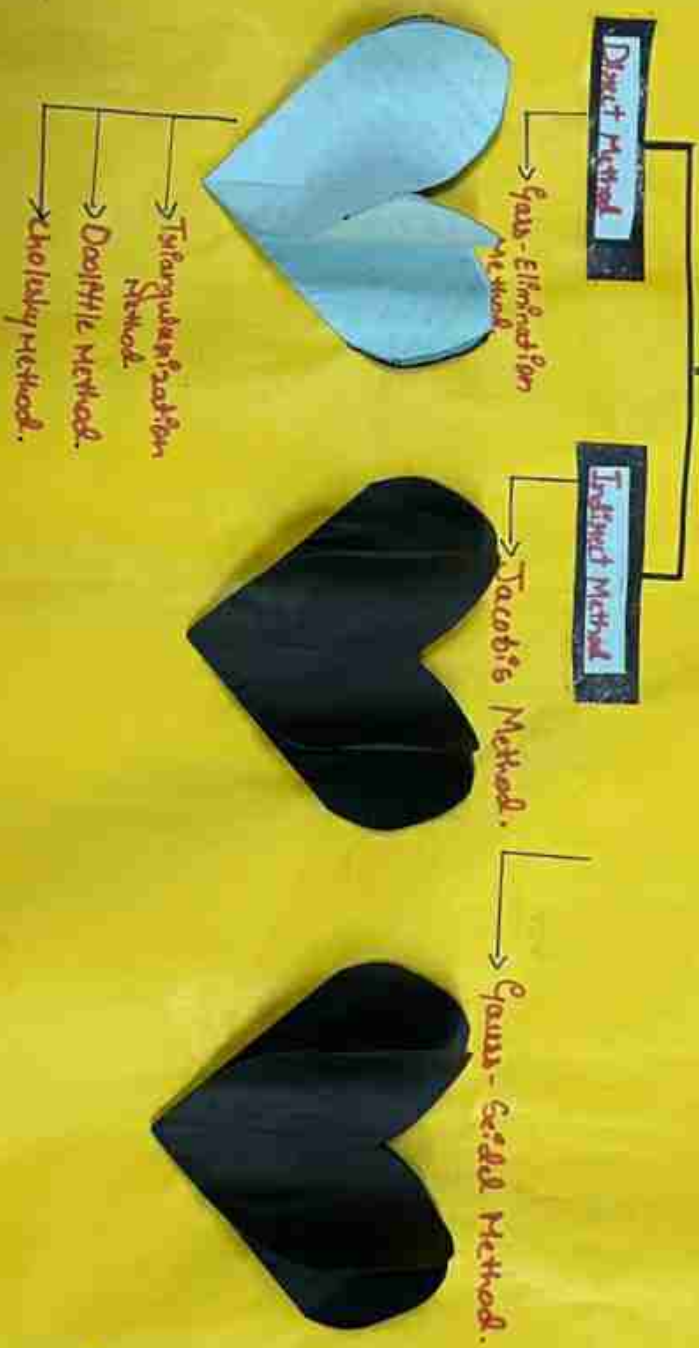
Handwritten notes in Hindi at the bottom left corner of the page.

NSM UNIT-3 SYSTEM OF LINEAR EQUATION

Method, "Cramer's rule" and "Matrix methods". These methods are used when the linear equation are very less in number. When the number of unknown in the system is large, the methods become tedious. To solve the such types of equation we use various other numerical methods which are suited for computer operations.

These numerical methods have 2 types

Digitation :- 5 for we have studied the methods of solving linear equation



Gyisu Ka. N.
Kumuda. M.
Ranyo. M. G.
Ranyo. N.

FLOATING POINT REPRESENTATION

The secant method converges for the approximated root similar to Newton-Raphson Method

Secant Method

$$x_{n+1} = x_n - \left(\frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} \right) f(x_n)$$

Newton's Method

The previous known method makes it also easy for Newton-Raphson Iteration

$$x_{n+1} = x_n - h$$

$$h = \frac{f(x_n)}{f'(x_n)}$$

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

Introduction: The aim of Numerical Method is to provide efficient methods for obtaining numerical answers to such problems with the above advent of the modern high speed electronic digital computer, our main concern is to provide numerical methods, efficient and suitable in different cases of higher mathematics.

Bisection Method

It is also called Interval Halving Method

$$x_p = \frac{a+b}{2}$$

Note: The bisection method requires large number of iteration to achieve a reasonable degree of accuracy for the root

Normalized floating point

In the decimal system any real number (which has any) can be represented in normalized floating point form as

$$x = \pm 0.1d_1d_2 \dots \times 10^E$$

$$E = \pm 02, 03, 04, \dots$$

$$d_i = 0, 1, 2, 3, \dots, 9$$

Range of sign \rightarrow -32768 to 32767
 Range of floating point \rightarrow 10^{-38} to 10^{38}

Subtract the floating point numbers

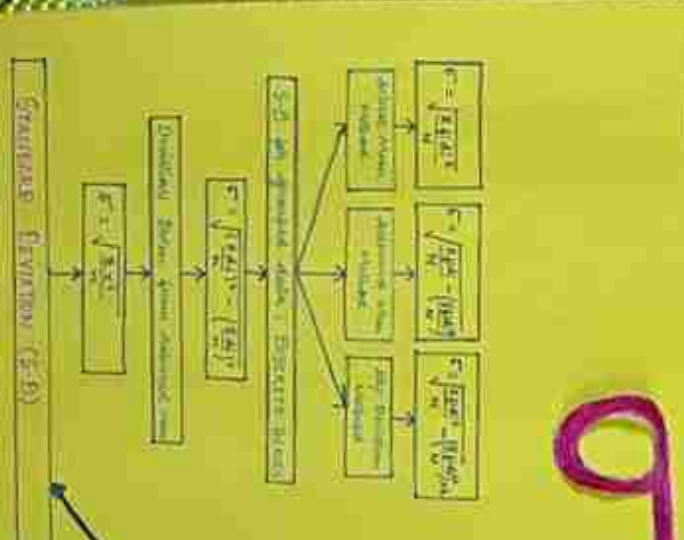
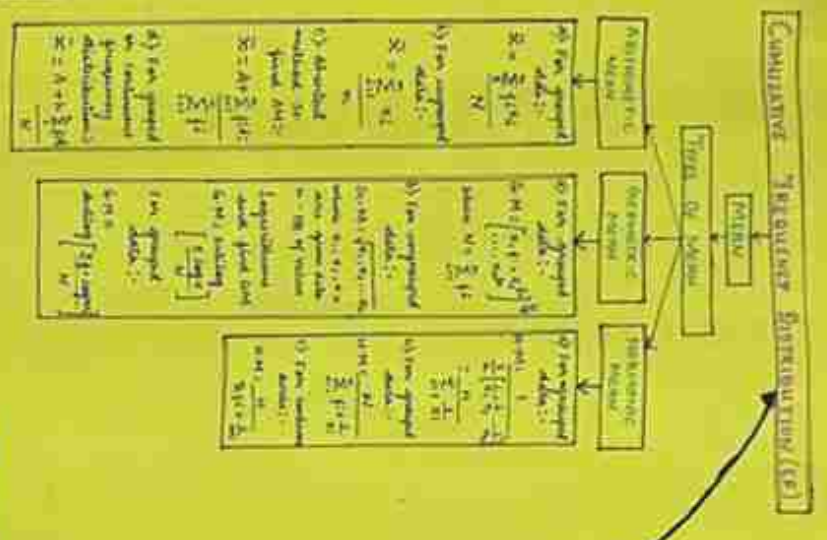
$$\begin{array}{r} 1) \ 0.36153117 \times 10^3 \\ - 0.561532546 \times 10^2 \\ \hline 0.80011701 \times 10^2 \end{array}$$

Single Precision

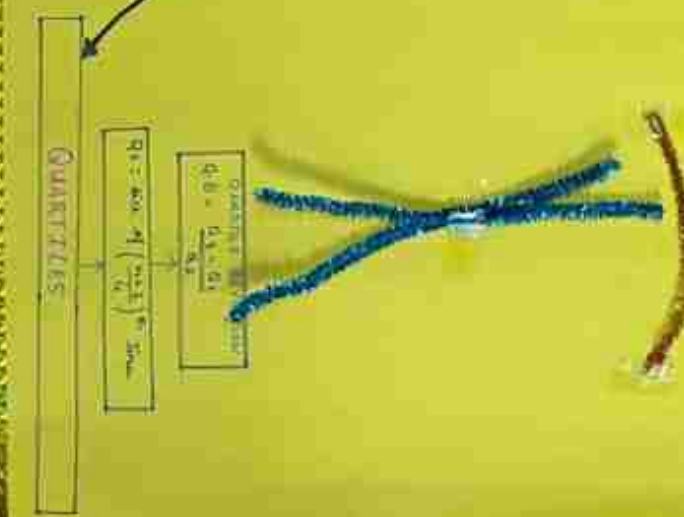
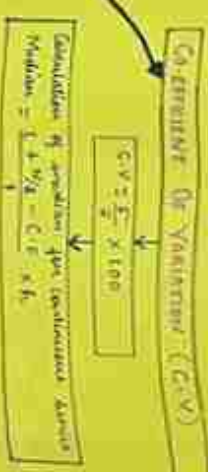
It is $(10)^8 \times 24$ (signed-bit) $\times (1-4)$ (significand)
 Two single precision $e = \pm 127$
 Signs bit : 0 - positive
 1 - negative
 Represent the numbers $\begin{array}{|c|c|c|} \hline 5 & 2 & 1 \\ \hline 1 & 8 & 25 \\ \hline \end{array}$
 $= 52.325$

Double Precision

$(10)^8 \times 52$ (signed-bit) $\times (1-8)$ (significand)
 Two double precision $e = \pm 1023$
 Signs bit : 0 - positive
 1 - negative
 Represent the numbers $\begin{array}{|c|c|c|} \hline 5 & 2 & 1 \\ \hline 1 & 15 & 53 \\ \hline \end{array}$
 $= 64.535$



STATISTICAL METHOD



UP
PR
BY
BSZ

System Of Linear Equations

Introduction:

Linear Equations can be solved in different methods. When there are two different Equations we can solve by using Graphical method or matrix method or Simultaneous linear Equation methods. When there are three different Equations we can solve by using Gauss Elimination method, Gauss-Jacobi method or Jacobi method.

Gauss Elimination Method

This method is illustrated by considering three independent Eqn's in three unknowns. This method is based on elimination of unknowns by combining Eqn's such that the 'n' Eqn's in 'n' unknowns are reduced to an equivalent upper triangular system which can be solved by back substitution.

Let us consider

$$\begin{cases} a_1x + b_1y + c_1z = d_1 \\ a_2x + b_2y + c_2z = d_2 \\ a_3x + b_3y + c_3z = d_3 \end{cases} \dots (1)$$

The system (1) is equivalent to the matrix Equation $AX = B$ where

$$A = \begin{bmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{bmatrix}, \quad X = \begin{bmatrix} x \\ y \\ z \end{bmatrix}, \quad B = \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix}$$

The method aims in reducing the coefficient matrix A , to upper triangular matrix

$$[A:B] = \begin{bmatrix} a_1 & b_1 & c_1 & d_1 \\ a_2 & b_2 & c_2 & d_2 \\ a_3 & b_3 & c_3 & d_3 \end{bmatrix}$$

Equations

Step 1: we use element $(1,1)$ to make the elements a_2, a_3 zero by elementary row transformation.

$$[A:B] = \begin{bmatrix} a_1 & b_1 & c_1 & d_1 \\ 0 & b_2 & c_2 & d_2 \\ 0 & b_3 & c_3 & d_3 \end{bmatrix} \dots (2)$$

Step 2: we use element b_2 to make elements b_3 zero by elementary row transformation.

$$[A:B] = \begin{bmatrix} a_1 & b_1 & c_1 & d_1 \\ 0 & b_2 & c_2 & d_2 \\ 0 & 0 & c_3 & d_3 \end{bmatrix} \dots (3)$$

From Eqn (3) we get

$$\begin{cases} a_1x + b_1y + c_1z = d_1 \\ b_2y + c_2z = d_2 \\ c_3z = d_3 \end{cases}$$

we get z from the last Eqn & back substitution we get x, y of equation.

Gauss-Jacobi method

Consider the system of Eqn's

$$\begin{cases} a_1x + b_1y + c_1z = d_1 \\ a_2x + b_2y + c_2z = d_2 \\ a_3x + b_3y + c_3z = d_3 \end{cases}$$

To solve in this method we may have to rearrange the given system of Eqn's to meet the requirement.

We have $x = \frac{1}{a_1}(d_1 - b_1y - c_1z) \dots (1)$

$$y = \frac{1}{b_2}(d_2 - a_2x - c_2z) \dots (2)$$

$$z = \frac{1}{c_3}(d_3 - a_3x - b_3z) \dots (3)$$

Let $x = x_1, y = y_1, z = z_1$ using these values (first iterative) in

$$x_1 = \frac{1}{a_1}(d_1 - b_1y_1 - c_1z_1), \quad y_1 = \frac{1}{b_2}(d_2 - a_2x_1 - c_2z_1), \quad z_1 = \frac{1}{c_3}(d_3 - a_3x_1 - b_3y_1)$$

$$x_2 = \frac{1}{a_1}(d_1 - a_2x_1 - b_3y_1)$$

Similarly the second iterative values are

$$x_2 = \frac{1}{a_1}(d_1 - b_1y_2 - c_1z_2)$$

$$y_2 = \frac{1}{b_2}(d_2 - a_2x_2 - c_2z_2)$$

$$z_2 = \frac{1}{c_3}(d_3 - a_3x_2 - b_3y_2)$$

The above process is repeated until two consecutive iterative values are same.

Gauss-Seidel method

It is the modification of Jacobi's method. Consider the Eqn's

$$\begin{cases} a_1x + b_1y + c_1z = d_1 \\ a_2x + b_2y + c_2z = d_2 \\ a_3x + b_3y + c_3z = d_3 \end{cases}$$

Let, $x = \frac{1}{a_1}(d_1 - b_1y - c_1z) \dots (1)$

$$y = \frac{1}{b_2}(d_2 - a_2x - c_2z) \dots (2)$$

$$z = \frac{1}{c_3}(d_3 - a_3x - b_3z) \dots (3)$$

For initial values let $x = x_0, y = y_0, z = z_0$

$$x_1 = \frac{1}{a_1}(d_1 - b_1y_0 - c_1z_0)$$

$$y_1 = \frac{1}{b_2}(d_2 - a_2x_1 - c_2z_0)$$

$$z_1 = \frac{1}{c_3}(d_3 - a_3x_1 - b_3y_1)$$

$$x_2 = \frac{1}{a_1}(d_1 - b_1y_1 - c_1z_1)$$

$$y_2 = \frac{1}{b_2}(d_2 - a_2x_2 - c_2z_1)$$

$$z_2 = \frac{1}{c_3}(d_3 - a_3x_2 - b_3y_2)$$

The same process is repeated until the same values of x, y, z are obtained.

Statistical methods

I) Continuous series

$$G.M = \left(\frac{\sum f x^k}{\sum f} \right)^{1/k}$$

III Harmonic mean (H.M)

1) Individual series

$$H.M = \frac{N}{\sum \frac{1}{x}}$$

where N = Total no. of observations
 x = Value of variable

2) Discrete series

$$H.M = \frac{N}{\sum f/x}$$

3) Continuous series

$$H.M = \frac{N}{\sum f/x}$$

Mean

1) Sturgeson's test
 (a) The number of subdivisions is odd then median is the middle value else the value here is obtained by taking the am of the middle two

2) Ogive series

Median (M) = Size of $(\frac{N}{2})^{th}$ item

3) Step deviation series

$$M = \text{value of } (\frac{N}{2})^{th} \text{ item}$$

Median (M) = $L + \frac{\frac{N}{2} - C}{f}$

L is lower limit
 f is class frequency
 C is cumulative frequency

Arithmetic mean

$$\bar{x} = \frac{\sum x}{N}$$

where x = Arithmetic mean of variable
 N = Sum of values of variable
 x = Value of variable
 N = Total no. of observations of item

Direct method

Original ungrouped data	Step deviation	Direct method
$\bar{x} = \frac{\sum x}{N}$	$\bar{x} = \frac{\sum f d}{N}$	$\bar{x} = \frac{\sum f x}{N}$
	L = starting point of class f = frequency d = deviation from L	L = starting point of class f = frequency x = value of class

Short cut method

Table of frequency	Step deviation	Direct method
$\bar{x} = \frac{\sum f x}{N}$	$\bar{x} = \frac{\sum f d}{N}$	$\bar{x} = \frac{\sum f x}{N}$
L = starting point of class f = frequency x = value of class	L = starting point of class f = frequency d = deviation from L	L = starting point of class f = frequency x = value of class

Step deviation method

Table of frequency	Step deviation	Direct method
$\bar{x} = \frac{\sum f x}{N}$	$\bar{x} = \frac{\sum f d}{N}$	$\bar{x} = \frac{\sum f x}{N}$
L = starting point of class f = frequency x = value of class	L = starting point of class f = frequency d = deviation from L	L = starting point of class f = frequency x = value of class

II Assumed mean (A.M)

1) Individual series

$$\bar{x} = A + \frac{\sum f d}{N}$$

2) Discrete series

$$\bar{x} = A + \frac{\sum f d}{N}$$

A = Assumed mean
 f = frequency
 d = deviation from A

Quantiles

1) For discrete

$$Q_1 = \text{Size of } (\frac{N}{4})^{th} \text{ item}$$

$$Q_3 = \text{Size of } 3(\frac{N}{4})^{th} \text{ item}$$

2) For continuous

$$Q_1 = \text{Size of } (\frac{N}{4})^{th} \text{ item}$$

$$Q_3 = \text{Size of } 3(\frac{N}{4})^{th} \text{ item}$$

$$Q_2 = \text{Median}$$

1) For individual series

The term is arranged in ascending order (or descending order) and the value that is present in the middle term is known as individual mean

2) For discrete

Here the mode is more frequent than any other value. The value that is present in the middle term is known as individual mean

3) Continuous

where $f = \frac{1}{2} (f_1 + f_2)$
 L = lower limit
 f = frequency
 h = interval or width
 N = total no.

Relation between quartiles and median

$$Q_3 - Q_1 = 3(M - Q_2)$$

STANDARD DEVIATION

$$\sigma = \sqrt{\frac{\sum f d^2}{N}}$$

Assumed mean

$$\sigma = \sqrt{\frac{\sum f d^2}{N}}$$

$$\sigma = \sqrt{\frac{\sum f x^2}{N} - \left(\frac{\sum f x}{N} \right)^2}$$

Discrete series

$$\sigma = \sqrt{\frac{\sum f x^2}{N} - \left(\frac{\sum f x}{N} \right)^2}$$

Assumed mean

$$\sigma = \sqrt{\frac{\sum f d^2}{N}}$$

Step deviation

$$\sigma = \sqrt{\frac{\sum f d^2}{N}}$$

Continuous

$$\sigma = \sqrt{\frac{\sum f d^2}{N}}$$

Assumed mean

$$\sigma = \sqrt{\frac{\sum f d^2}{N}}$$

Step deviation

$$\sigma = \sqrt{\frac{\sum f d^2}{N}}$$

Coefficient of variation

$$C.V = \frac{\sigma}{\bar{x}} \times 100$$

Skewness (and mean)

Skewness is the measure of the asymmetry of the distribution. It is denoted by β_1 .
 $\beta_1 = \frac{\sum f x^3}{N \bar{x}^3}$

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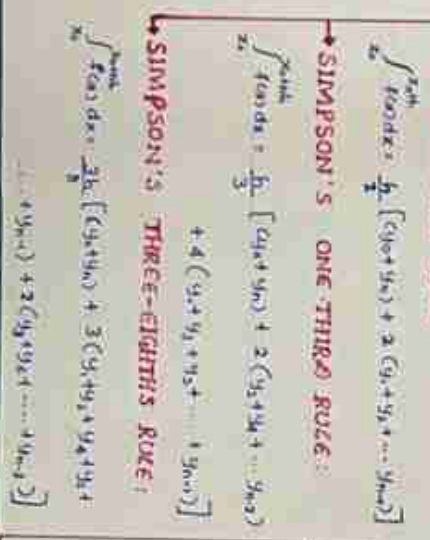
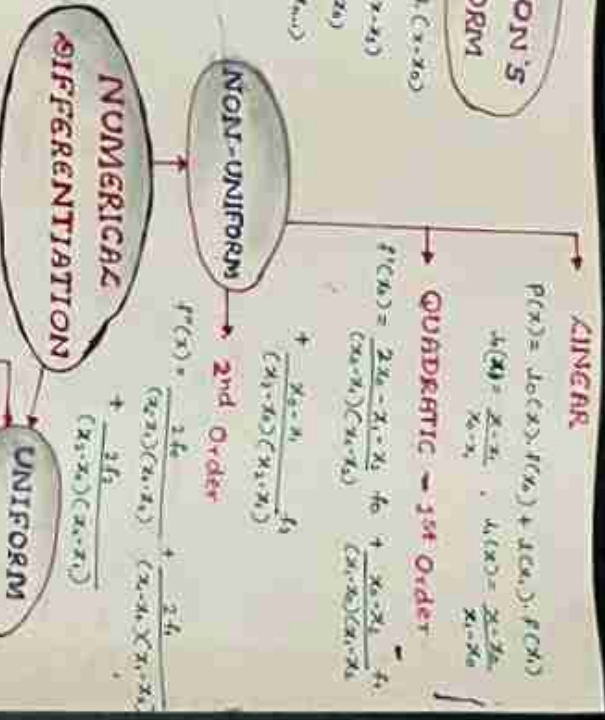
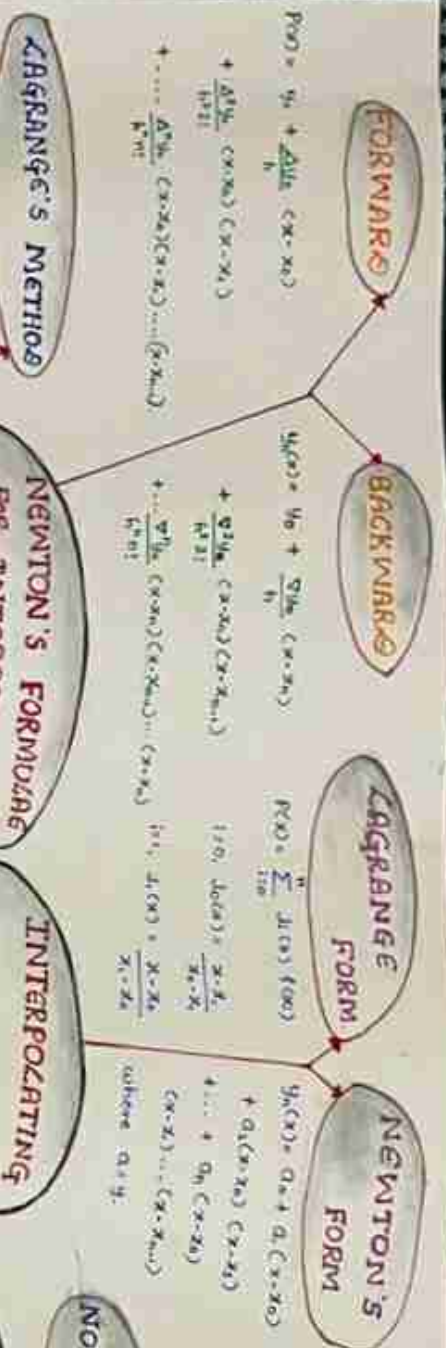
Skewness (and mean)

$$\beta_1 = \frac{\sum f x^3}{N \bar{x}^3}$$

Skewness (and mean)

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NUMERICAL DIFFERENTIATION & INTERPOLATION

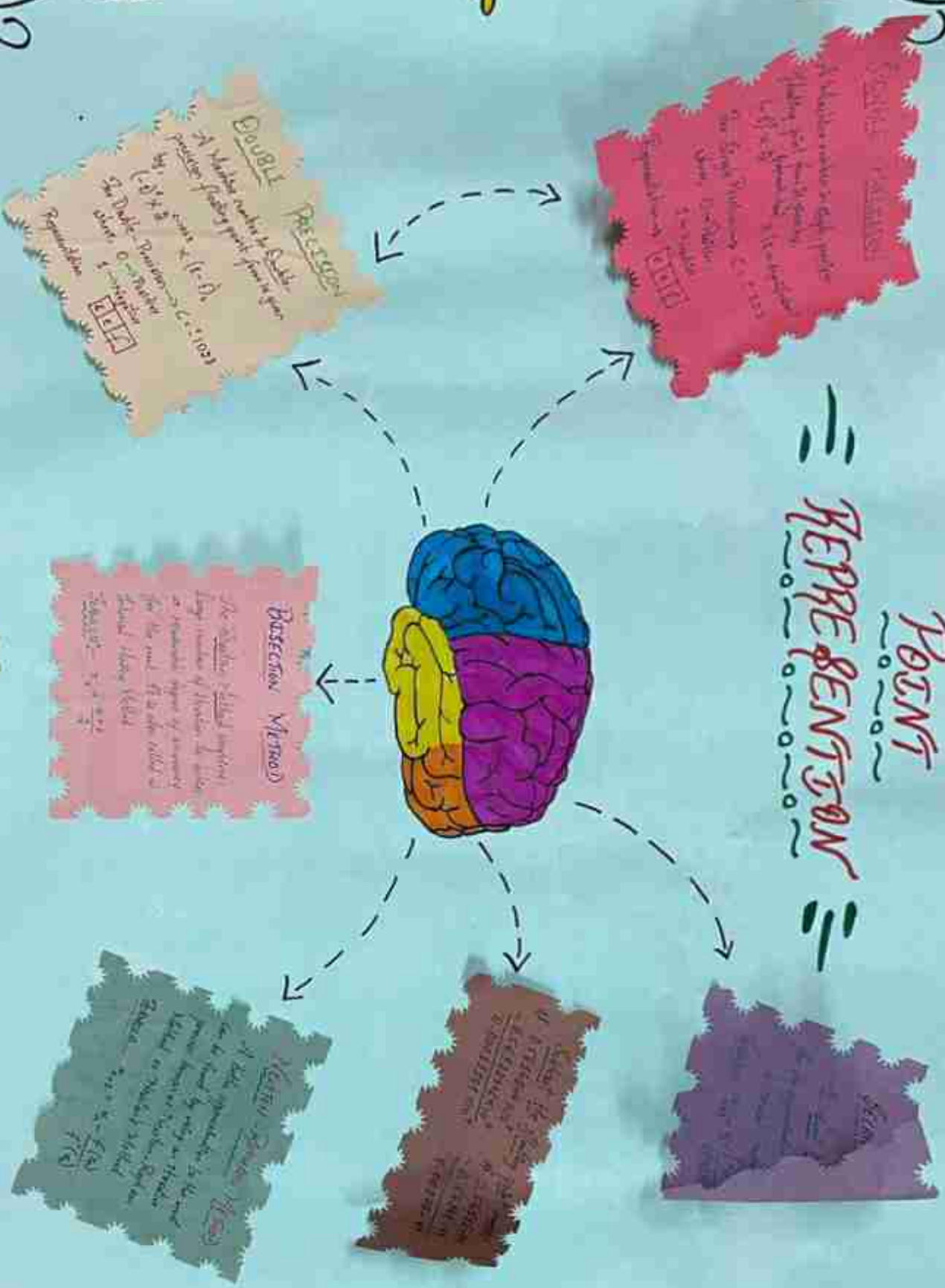


x	y	Δy	$\Delta^2 y$	$\Delta^3 y$	$\Delta^4 y$	x	y	Δy	$\Delta^2 y$	$\Delta^3 y$	$\Delta^4 y$
x_0	y_0	Δy_0	$\Delta^2 y_0$	$\Delta^3 y_0$	$\Delta^4 y_0$	x_1	y_1	Δy_1	$\Delta^2 y_1$	$\Delta^3 y_1$	$\Delta^4 y_1$
x_2	y_2	Δy_2	$\Delta^2 y_2$	$\Delta^3 y_2$	$\Delta^4 y_2$	x_2	y_2	Δy_2	$\Delta^2 y_2$	$\Delta^3 y_2$	$\Delta^4 y_2$
x_3	y_3	Δy_3	$\Delta^2 y_3$	$\Delta^3 y_3$	$\Delta^4 y_3$	x_3	y_3	Δy_3	$\Delta^2 y_3$	$\Delta^3 y_3$	$\Delta^4 y_3$
x_4	y_4	Δy_4	$\Delta^2 y_4$	$\Delta^3 y_4$	$\Delta^4 y_4$	x_4	y_4	Δy_4	$\Delta^2 y_4$	$\Delta^3 y_4$	$\Delta^4 y_4$
x_5	y_5	Δy_5	$\Delta^2 y_5$	$\Delta^3 y_5$	$\Delta^4 y_5$	x_5	y_5	Δy_5	$\Delta^2 y_5$	$\Delta^3 y_5$	$\Delta^4 y_5$
x_6	y_6	Δy_6	$\Delta^2 y_6$	$\Delta^3 y_6$	$\Delta^4 y_6$	x_6	y_6	Δy_6	$\Delta^2 y_6$	$\Delta^3 y_6$	$\Delta^4 y_6$
x_7	y_7	Δy_7	$\Delta^2 y_7$	$\Delta^3 y_7$	$\Delta^4 y_7$	x_7	y_7	Δy_7	$\Delta^2 y_7$	$\Delta^3 y_7$	$\Delta^4 y_7$
x_8	y_8	Δy_8	$\Delta^2 y_8$	$\Delta^3 y_8$	$\Delta^4 y_8$	x_8	y_8	Δy_8	$\Delta^2 y_8$	$\Delta^3 y_8$	$\Delta^4 y_8$
x_9	y_9	Δy_9	$\Delta^2 y_9$	$\Delta^3 y_9$	$\Delta^4 y_9$	x_9	y_9	Δy_9	$\Delta^2 y_9$	$\Delta^3 y_9$	$\Delta^4 y_9$

x	$f(x)$	$\Delta f(x)$	$\Delta^2 f(x)$	$\Delta^3 f(x)$
x_0	$f(x_0)$	$\frac{f(x_1) - f(x_0)}{x_1 - x_0}$	$\frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}$	$\frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1} + \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_1	$f(x_1)$	$\frac{f(x_2) - f(x_1)}{x_2 - x_1}$	$\frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1}$	$\frac{f(x_4) - f(x_3)}{x_4 - x_3} - \frac{f(x_3) - f(x_2)}{x_3 - x_2} + \frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_2	$f(x_2)$	$\frac{f(x_3) - f(x_2)}{x_3 - x_2}$	$\frac{f(x_4) - f(x_3)}{x_4 - x_3} - \frac{f(x_3) - f(x_2)}{x_3 - x_2}$	$\frac{f(x_5) - f(x_4)}{x_5 - x_4} - \frac{f(x_4) - f(x_3)}{x_4 - x_3} + \frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1} + \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_3	$f(x_3)$	$\frac{f(x_4) - f(x_3)}{x_4 - x_3}$	$\frac{f(x_5) - f(x_4)}{x_5 - x_4} - \frac{f(x_4) - f(x_3)}{x_4 - x_3}$	$\frac{f(x_6) - f(x_5)}{x_6 - x_5} - \frac{f(x_5) - f(x_4)}{x_5 - x_4} + \frac{f(x_4) - f(x_3)}{x_4 - x_3} - \frac{f(x_3) - f(x_2)}{x_3 - x_2} + \frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_4	$f(x_4)$	$\frac{f(x_5) - f(x_4)}{x_5 - x_4}$	$\frac{f(x_6) - f(x_5)}{x_6 - x_5} - \frac{f(x_5) - f(x_4)}{x_5 - x_4}$	$\frac{f(x_7) - f(x_6)}{x_7 - x_6} - \frac{f(x_6) - f(x_5)}{x_6 - x_5} + \frac{f(x_5) - f(x_4)}{x_5 - x_4} - \frac{f(x_4) - f(x_3)}{x_4 - x_3} + \frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1} + \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_5	$f(x_5)$	$\frac{f(x_6) - f(x_5)}{x_6 - x_5}$	$\frac{f(x_7) - f(x_6)}{x_7 - x_6} - \frac{f(x_6) - f(x_5)}{x_6 - x_5}$	$\frac{f(x_8) - f(x_7)}{x_8 - x_7} - \frac{f(x_7) - f(x_6)}{x_7 - x_6} + \frac{f(x_6) - f(x_5)}{x_6 - x_5} - \frac{f(x_5) - f(x_4)}{x_5 - x_4} + \frac{f(x_4) - f(x_3)}{x_4 - x_3} - \frac{f(x_3) - f(x_2)}{x_3 - x_2} + \frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_6	$f(x_6)$	$\frac{f(x_7) - f(x_6)}{x_7 - x_6}$	$\frac{f(x_8) - f(x_7)}{x_8 - x_7} - \frac{f(x_7) - f(x_6)}{x_7 - x_6}$	$\frac{f(x_9) - f(x_8)}{x_9 - x_8} - \frac{f(x_8) - f(x_7)}{x_8 - x_7} + \frac{f(x_7) - f(x_6)}{x_7 - x_6} - \frac{f(x_6) - f(x_5)}{x_6 - x_5} + \frac{f(x_5) - f(x_4)}{x_5 - x_4} - \frac{f(x_4) - f(x_3)}{x_4 - x_3} + \frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1} + \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_7	$f(x_7)$	$\frac{f(x_8) - f(x_7)}{x_8 - x_7}$	$\frac{f(x_9) - f(x_8)}{x_9 - x_8} - \frac{f(x_8) - f(x_7)}{x_8 - x_7}$	$\frac{f(x_{10}) - f(x_9)}{x_{10} - x_9} - \frac{f(x_9) - f(x_8)}{x_9 - x_8} + \frac{f(x_8) - f(x_7)}{x_8 - x_7} - \frac{f(x_7) - f(x_6)}{x_7 - x_6} + \frac{f(x_6) - f(x_5)}{x_6 - x_5} - \frac{f(x_5) - f(x_4)}{x_5 - x_4} + \frac{f(x_4) - f(x_3)}{x_4 - x_3} - \frac{f(x_3) - f(x_2)}{x_3 - x_2} + \frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_8	$f(x_8)$	$\frac{f(x_9) - f(x_8)}{x_9 - x_8}$	$\frac{f(x_{10}) - f(x_9)}{x_{10} - x_9} - \frac{f(x_9) - f(x_8)}{x_9 - x_8}$	$\frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}} - \frac{f(x_{10}) - f(x_9)}{x_{10} - x_9} + \frac{f(x_9) - f(x_8)}{x_9 - x_8} - \frac{f(x_8) - f(x_7)}{x_8 - x_7} + \frac{f(x_7) - f(x_6)}{x_7 - x_6} - \frac{f(x_6) - f(x_5)}{x_6 - x_5} + \frac{f(x_5) - f(x_4)}{x_5 - x_4} - \frac{f(x_4) - f(x_3)}{x_4 - x_3} + \frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1} + \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_9	$f(x_9)$	$\frac{f(x_{10}) - f(x_9)}{x_{10} - x_9}$	$\frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}} - \frac{f(x_{10}) - f(x_9)}{x_{10} - x_9}$	$\frac{f(x_{12}) - f(x_{11})}{x_{12} - x_{11}} - \frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}} + \frac{f(x_{10}) - f(x_9)}{x_{10} - x_9} - \frac{f(x_9) - f(x_8)}{x_9 - x_8} + \frac{f(x_8) - f(x_7)}{x_8 - x_7} - \frac{f(x_7) - f(x_6)}{x_7 - x_6} + \frac{f(x_6) - f(x_5)}{x_6 - x_5} - \frac{f(x_5) - f(x_4)}{x_5 - x_4} + \frac{f(x_4) - f(x_3)}{x_4 - x_3} - \frac{f(x_3) - f(x_2)}{x_3 - x_2} + \frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_{10}	$f(x_{10})$	$\frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}}$	$\frac{f(x_{12}) - f(x_{11})}{x_{12} - x_{11}} - \frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}}$	$\frac{f(x_{13}) - f(x_{12})}{x_{13} - x_{12}} - \frac{f(x_{12}) - f(x_{11})}{x_{12} - x_{11}} + \frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}} - \frac{f(x_{10}) - f(x_9)}{x_{10} - x_9} + \frac{f(x_9) - f(x_8)}{x_9 - x_8} - \frac{f(x_8) - f(x_7)}{x_8 - x_7} + \frac{f(x_7) - f(x_6)}{x_7 - x_6} - \frac{f(x_6) - f(x_5)}{x_6 - x_5} + \frac{f(x_5) - f(x_4)}{x_5 - x_4} - \frac{f(x_4) - f(x_3)}{x_4 - x_3} + \frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1} + \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_{11}	$f(x_{11})$	$\frac{f(x_{12}) - f(x_{11})}{x_{12} - x_{11}}$	$\frac{f(x_{13}) - f(x_{12})}{x_{13} - x_{12}} - \frac{f(x_{12}) - f(x_{11})}{x_{12} - x_{11}}$	$\frac{f(x_{14}) - f(x_{13})}{x_{14} - x_{13}} - \frac{f(x_{13}) - f(x_{12})}{x_{13} - x_{12}} + \frac{f(x_{12}) - f(x_{11})}{x_{12} - x_{11}} - \frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}} + \frac{f(x_{10}) - f(x_9)}{x_{10} - x_9} - \frac{f(x_9) - f(x_8)}{x_9 - x_8} + \frac{f(x_8) - f(x_7)}{x_8 - x_7} - \frac{f(x_7) - f(x_6)}{x_7 - x_6} + \frac{f(x_6) - f(x_5)}{x_6 - x_5} - \frac{f(x_5) - f(x_4)}{x_5 - x_4} + \frac{f(x_4) - f(x_3)}{x_4 - x_3} - \frac{f(x_3) - f(x_2)}{x_3 - x_2} + \frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_{12}	$f(x_{12})$	$\frac{f(x_{13}) - f(x_{12})}{x_{13} - x_{12}}$	$\frac{f(x_{14}) - f(x_{13})}{x_{14} - x_{13}} - \frac{f(x_{13}) - f(x_{12})}{x_{13} - x_{12}}$	$\frac{f(x_{15}) - f(x_{14})}{x_{15} - x_{14}} - \frac{f(x_{14}) - f(x_{13})}{x_{14} - x_{13}} + \frac{f(x_{13}) - f(x_{12})}{x_{13} - x_{12}} - \frac{f(x_{12}) - f(x_{11})}{x_{12} - x_{11}} + \frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}} - \frac{f(x_{10}) - f(x_9)}{x_{10} - x_9} + \frac{f(x_9) - f(x_8)}{x_9 - x_8} - \frac{f(x_8) - f(x_7)}{x_8 - x_7} + \frac{f(x_7) - f(x_6)}{x_7 - x_6} - \frac{f(x_6) - f(x_5)}{x_6 - x_5} + \frac{f(x_5) - f(x_4)}{x_5 - x_4} - \frac{f(x_4) - f(x_3)}{x_4 - x_3} + \frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1} + \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_{13}	$f(x_{13})$	$\frac{f(x_{14}) - f(x_{13})}{x_{14} - x_{13}}$	$\frac{f(x_{15}) - f(x_{14})}{x_{15} - x_{14}} - \frac{f(x_{14}) - f(x_{13})}{x_{14} - x_{13}}$	$\frac{f(x_{16}) - f(x_{15})}{x_{16} - x_{15}} - \frac{f(x_{15}) - f(x_{14})}{x_{15} - x_{14}} + \frac{f(x_{14}) - f(x_{13})}{x_{14} - x_{13}} - \frac{f(x_{13}) - f(x_{12})}{x_{13} - x_{12}} + \frac{f(x_{12}) - f(x_{11})}{x_{12} - x_{11}} - \frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}} + \frac{f(x_{10}) - f(x_9)}{x_{10} - x_9} - \frac{f(x_9) - f(x_8)}{x_9 - x_8} + \frac{f(x_8) - f(x_7)}{x_8 - x_7} - \frac{f(x_7) - f(x_6)}{x_7 - x_6} + \frac{f(x_6) - f(x_5)}{x_6 - x_5} - \frac{f(x_5) - f(x_4)}{x_5 - x_4} + \frac{f(x_4) - f(x_3)}{x_4 - x_3} - \frac{f(x_3) - f(x_2)}{x_3 - x_2} + \frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_{14}	$f(x_{14})$	$\frac{f(x_{15}) - f(x_{14})}{x_{15} - x_{14}}$	$\frac{f(x_{16}) - f(x_{15})}{x_{16} - x_{15}} - \frac{f(x_{15}) - f(x_{14})}{x_{15} - x_{14}}$	$\frac{f(x_{17}) - f(x_{16})}{x_{17} - x_{16}} - \frac{f(x_{16}) - f(x_{15})}{x_{16} - x_{15}} + \frac{f(x_{15}) - f(x_{14})}{x_{15} - x_{14}} - \frac{f(x_{14}) - f(x_{13})}{x_{14} - x_{13}} + \frac{f(x_{13}) - f(x_{12})}{x_{13} - x_{12}} - \frac{f(x_{12}) - f(x_{11})}{x_{12} - x_{11}} + \frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}} - \frac{f(x_{10}) - f(x_9)}{x_{10} - x_9} + \frac{f(x_9) - f(x_8)}{x_9 - x_8} - \frac{f(x_8) - f(x_7)}{x_8 - x_7} + \frac{f(x_7) - f(x_6)}{x_7 - x_6} - \frac{f(x_6) - f(x_5)}{x_6 - x_5} + \frac{f(x_5) - f(x_4)}{x_5 - x_4} - \frac{f(x_4) - f(x_3)}{x_4 - x_3} + \frac{f(x_3) - f(x_2)}{x_3 - x_2} - \frac{f(x_2) - f(x_1)}{x_2 - x_1} + \frac{f(x_1) - f(x_0)}{x_1 - x_0}$
x_{15}	$f(x_{15})$	$\frac{f(x_{16}) - f(x_{15})}{x_{16} - x_{15}}$	$\frac{f(x_{17}) - f(x_{16})}{x_{17} - x_{16}} - \frac{f(x_{16}) - f(x_{15})}{x_{16} - x_{15}}$	$\frac{f(x_{18}) - f(x_{17})}{x_{18} - x_{17}} - \frac{f(x_{17}) - f(x_{16})}{x_{17} - x_{16}} + \frac{f(x_{16}) - f(x_{15})}{x_{16} - x_{15}} - \frac{f(x_{15}) - f(x_{14})}{x_{15} - x_{14}} + \frac{f(x_{14}) - f(x_{13})}{x_{14} - x_{13}} - \frac{f(x_{13}) - f(x_{12})}{x_{13} - x_{12}} + \frac{f(x_{12}) - f(x_{11})}{x_{12} - x_{11}} - \frac{f(x_{11}) - f(x_{10})}{x_{11} - x_{10}} + \frac{f(x_{10}) - f(x_9)}{x_{10} - x_9} - \frac{f(x_9) - f(x_8)}{x_9 - x_8} + \frac{f(x_8) - f(x_7)}{x_8 - x_7} - \frac{f(x_7) - f(x_6)}{x_7 - x_6} + \frac{f(x_6) - f(x_5)}{x_6 - x_5} - \frac{f(x_5) - f(x_4)}{x_5 - x_4} + \frac{f(x_4) - f(x_3)}{x_4 - x_3} - \frac{f(x_3) - f(x_2)}{x_3 - x_2} + \frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}$

FLOATING POINT

REPRESENTATION



DOUBLE PRECISION
 A double-precision floating-point number is a number that is represented by two single-precision floating-point numbers.
 It is defined by IEEE 754-2008.
 It has a 64-bit format.
 It consists of a 1-bit sign, an 11-bit exponent, and a 52-bit mantissa.
 It is used for high-precision calculations.

FLOATING POINT REPRESENTATION
 A floating-point number is a number that is represented by a sign, an exponent, and a mantissa.
 It is defined by IEEE 754-2008.
 It has a 32-bit format.
 It consists of a 1-bit sign, an 8-bit exponent, and a 23-bit mantissa.
 It is used for most calculations.

BASECATION METHOD
 The basecacion method is a method for converting a number from one base to another.
 It is defined by IEEE 754-2008.
 It is used for converting numbers between different bases.

IEEE 754 STANDARD
 The IEEE 754 standard is a standard for floating-point arithmetic.
 It is defined by IEEE 754-2008.
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Mitch
 Vashishth



Polynomial (18) 200
 $f(x) = a_0 + a_1x + a_2x^2 + \dots + a_nx^n$
 $f(x_0) = a_0 + a_1x_0 + a_2x_0^2 + \dots + a_nx_0^n$
 $f(x_1) = a_0 + a_1x_1 + a_2x_1^2 + \dots + a_nx_1^n$
 \vdots
 $f(x_n) = a_0 + a_1x_n + a_2x_n^2 + \dots + a_nx_n^n$

Newton's general difference divided formula:-
 $f(x) = f(x_0) + (x-x_0)f'(x_0) + \frac{(x-x_0)^2}{2!}f''(x_0) + \dots$
 $f(x_1, x_2) + \dots + \frac{(x-x_0)(x-x_1)\dots(x-x_{n-1})}{n!}f^{(n)}(x_0)$

Interpolation & Differentiation

The divided difference table:

Argument	$f(x)$	$f'(x)$	$f''(x)$
x_0	$f(x_0)$	$f'(x_0)$	$f''(x_0)$
x_1	$f(x_1)$	$f'(x_1)$	$f''(x_1)$
x_2	$f(x_2)$	$f'(x_2)$	$f''(x_2)$
\vdots	\vdots	\vdots	\vdots
x_n	$f(x_n)$	$f'(x_n)$	$f''(x_n)$

	x_0	x_1	x_2	x_3	x_4
$f(x)$	f_0	f_1	f_2	f_3	f_4
$f'(x)$	f'_0	f'_1	f'_2	f'_3	f'_4
$f''(x)$	f''_0	f''_1	f''_2	f''_3	f''_4
$f'''(x)$	f'''_0	f'''_1	f'''_2	f'''_3	f'''_4

Newton's form:-
 $f_p(x) = q_0 + a_1(x-x_0) + a_2(x-x_0)(x-x_1) + \dots + a_n(x-x_0)\dots(x-x_{n-1})$
 when $a=4$

	x_0	x_1	x_2	x_3	x_4
$f(x)$	f_0	f_1	f_2	f_3	f_4
$f'(x)$	f'_0	f'_1	f'_2	f'_3	f'_4
$f''(x)$	f''_0	f''_1	f''_2	f''_3	f''_4
$f'''(x)$	f'''_0	f'''_1	f'''_2	f'''_3	f'''_4

Lagrange form:
 $f(x) = \sum_{i=0}^n L_i(x) f(x_i)$
 $L_0(x) = \frac{(x-x_1)(x-x_2)\dots(x-x_n)}{(x_0-x_1)(x_0-x_2)\dots(x_0-x_n)}$
 $L_1(x) = \frac{(x-x_0)(x-x_2)\dots(x-x_n)}{(x_1-x_0)(x_1-x_2)\dots(x_1-x_n)}$
 \vdots
 $L_n(x) = \frac{(x-x_0)(x-x_1)\dots(x-x_{n-1})}{(x_n-x_0)(x_n-x_1)\dots(x_n-x_{n-1})}$

Sturm's Interpolation:
Lagrange method:
 $f(x) = \frac{(x-x_1)(x-x_2)\dots(x-x_n)}{(x_0-x_1)(x_0-x_2)\dots(x_0-x_n)} f(x_0) + \dots + \frac{(x-x_0)(x-x_1)\dots(x-x_{n-1})}{(x_n-x_0)(x_n-x_1)\dots(x_n-x_{n-1})} f(x_n)$

Numerical differentiation:
Quadratic Interpolation \rightarrow 1st order:
 $f'(x) = \frac{f(x_1) - f(x_0)}{(x_1 - x_0)} + \frac{f(x_2) - f(x_1)}{(x_2 - x_1)} f$
2nd order:
 $f''(x) = \frac{2f_1 - f_0 - f_2}{(x_1 - x_0)(x_1 - x_2)} + \frac{2f_2 - f_1 - f_3}{(x_2 - x_1)(x_2 - x_3)}$



From memory:-
 Rishi-4
 Anshuman-3
 Gaurav-raddy-48