

GAUSSIAN ELEMINATION METHOD

Problem-1

Solve the following system of linear equations by Gaussian elimination method

$$10x_1 + x_2 - 5x_3 = 1$$

$$-20x_1 + 3x_2 + 20x_3 = 2$$

$$5x_1 + 3x_2 + 5x_3 = 6$$

Solⁿ -

a) Forward elimination:

$$10x_1 + x_2 - 5x_3 = 1 \quad \text{--- (1)}^0$$

$$-20x_1 + 3x_2 + 20x_3 = 2 \quad \text{--- (2)}^0$$

$$5x_1 + 3x_2 + 5x_3 = 6 \quad \text{--- (3)}^0$$

Elimination of the co-efficients of 'x₁' from the eqⁿ (2)⁰ & (3)⁰:-

$$(1)^1 = (1)^0 \longrightarrow 10x_1 + x_2 - 5x_3 = 1 \quad \text{--- (1)}^1$$

$$(2)^1 = (2)^0 - \left(\frac{-20}{10}\right) \times (1)^0 \longrightarrow 5x_2 + 10x_3 = 4 \quad \text{--- (2)}^1$$

$$(3)^1 = (3)^0 - \left(\frac{5}{10}\right) \times (1)^0 \longrightarrow 2.5x_2 + 7.5x_3 = 5.5 \quad \text{--- (3)}^1$$

Elimination of the co-efficient of 'x₂' from eqⁿ (3)¹:-

$$(1)^2 = (1)^1 \longrightarrow 10x_1 + x_2 - 5x_3 = 1 \quad \text{--- (1)}^2$$

$$(2)^2 = (2)^1 \longrightarrow 5x_2 + 10x_3 = 4 \quad \text{--- (2)}^2$$

$$(3)^2 = (3)^1 - \left(\frac{2.5}{5}\right) \times (2)^1 \longrightarrow 2.5x_3 = 3.5 \quad \text{--- (3)}^2$$

In matrix form we can write $UX = B$

$$\text{or } \begin{bmatrix} 10 & 1 & -5 \\ 0 & 5 & 10 \\ 0 & 0 & 2.5 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} 1 \\ 4 \\ 3.5 \end{Bmatrix}$$

b) Backward substitution:-

$$\text{From eqⁿ (3)}^2 \rightarrow x_3 = \frac{3.5}{2.5} = 1.4$$

$$\text{From eqⁿ (2)}^2 \rightarrow 5x_2 + 10 \times 1.4 = 4 \\ \text{or } x_2 = -2$$

$$\text{From eqⁿ (1)}^2 \rightarrow 10x_1 - 2 - 5 \times 1.4 = 1 \\ \text{or } x_1 = 1$$

The solution obtained :-

$$x_1 = 1$$

$$x_2 = -2$$

$$x_3 = 1.4$$

JACOBI ITERATION METHOD

General form of the i^{th} equation -

$$\sum_{j=1}^n a_{ij} x_j = b_i$$

From this we get $x_i = \frac{1}{a_{ii}} \left(b_i - \sum_{j=1}^{i-1} a_{ij} x_j - \sum_{j=i+1}^n a_{ij} x_j \right)$

By successive iteration we have

$$x_i^{(0)} \rightarrow x_i^{(1)} \rightarrow x_i^{(2)} \rightarrow \dots$$

$$x_i^{(1)} = \frac{1}{a_{ii}} \left[b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(0)} - \sum_{j=i+1}^n a_{ij} x_j^{(0)} \right]$$

$$x_i^{(k+1)} = \frac{1}{a_{ii}} \left[b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(k)} - \sum_{j=i+1}^n a_{ij} x_j^{(k)} \right]$$

By adding and subtracting we get,

$$x_i^{(k+1)} = x_i^{(k)} + \frac{1}{a_{ii}} \left[b_i - \sum_{j=1}^n a_{ij} x_j^{(k)} \right]$$

$$\text{or } x_i^{(k+1)} = x_i^{(k)} + \frac{R_i^{(k)}}{a_{ii}} \quad \left[R_i^{(k)} = b_i - \sum_{j=1}^n a_{ij} x_j^{(k)} \right]$$

Prob. - 6

Solve the following system of linear equations by using Jacobi iteration method.

$$16x_1 + 4x_2 + 8x_3 = 4$$

$$4x_1 + 5x_2 - 4x_3 = 2$$

$$8x_1 - 4x_2 + 22x_3 = 5$$

Soln.

Scaling the equations by the corresponding leading diagonal elements we can write the equations in the following form:

$$x_1 + 0.25x_2 + 0.5x_3 = 0.25$$

$$0.8x_1 + x_2 - 0.8x_3 = 0.40$$

$$0.3636x_1 - 0.1818x_2 + x_3 = 0.2273$$

$$\therefore x_1^{(k+1)} = 0.25 - 0.25 x_2^{(k)} - 0.5 x_3^{(k)}$$

$$x_2^{(k+1)} = 0.4 - 0.8 x_1^{(k)} + 0.8 x_3^{(k)}$$

$$x_3^{(k+1)} = 0.2273 - 0.3636 x_1^{(k)} + 0.1818 x_2^{(k)}$$

	Iteration numbers						
	1	2	3	4	5	6	7
x_1	1	-0.5	0.12725	-0.20001	-0.087458	-0.168776	-0.157031
x_2	1	0.4	0.8364	0.683656	0.826479	0.809416	0.86250
x_3	1	0.0455	0.48182	0.333089	0.424312	0.409354	0.43582

\therefore The solution obtained;

$$x_1 = -0.157031$$

$$x_2 = 0.86250$$

$$x_3 = 0.43582$$

GAUSS-SEIDEL METHOD

$$x_i^{(k+1)} = \frac{1}{a_{ii}} \left[b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(k+1)} - \sum_{j=i+1}^n a_{ij} x_j^{(k)} \right]$$

$$x_i^{(k+1)} = x_i^{(k)} + \frac{R_i^{(k)}}{a_{ii}} \quad \left[R_i^{(k)} = b_i - \sum_{j=1}^{i-1} a_{ij} x_j^{(k+1)} - \sum_{j=i+1}^n a_{ij} x_j^{(k)} \right]$$

Prob.-7

Solve the following system of linear equations by Gauss-seidel iteration method.

$$16x_1 + 4x_2 + 8x_3 = 4$$

$$4x_1 + 5x_2 - 4x_3 = 2$$

$$8x_1 - 4x_2 + 22x_3 = 5$$

Sol:-

Scaling the equation by the corresponding leading diagonal elements we can write the equations in the form:

$$x_1 + 0.25x_2 + 0.5x_3 = 0.25$$

$$0.8x_1 + x_2 - 0.8x_3 = 0.4$$

$$0.3636x_1 - 0.1818x_2 + x_3 = 0.2273$$

The equations can be written in the following form:

$$x_1^{(k+1)} = 0.25 - 0.25x_2^{(k)} - 0.5x_3^{(k)}$$

$$x_2^{(k+1)} = 0.4 - 0.8x_1^{(k+1)} + 0.8x_3^{(k)}$$

$$x_3^{(k+1)} = 0.2273 - 0.3636x_1^{(k+1)} + 0.1818x_2^{(k+1)}$$

	Iteration numbers						
	1	2	3	4	5	6	7
x_1	1	-0.5	-0.49999	-0.41816	-0.36907	-0.33366	-0.30885
x_2	1	1.6	1.35997	1.2596	1.18193	1.12802	1.09003
x_3	1	0.69998	0.65634	0.60834	0.57637	0.55369	0.53776

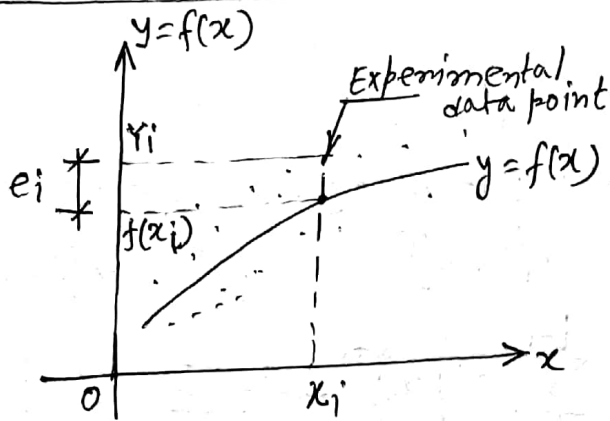
∴ The solution obtained:

$$x_1 = -0.30885$$

$$x_2 = 1.09003$$

$$x_3 = 0.53776$$

Least square method of curve-fitting or Regression



Let ~~n~~ ⁿ nos. of data points obtained from experiment.

$(x_i, y_i), i = 1, 2, \dots, n$

The error of approximation at $x = x_i$

$$e_i = y_i - f(x_i)$$

$$S = e_1^2 + e_2^2 + \dots + e_n^2$$

$$= [y_1 - f(x_1)]^2 + [y_2 - f(x_2)]^2 + \dots + [y_n - f(x_n)]^2$$

$$= \sum_{i=1}^n [y_i - f(x_i)]^2$$

principle. $i=1$

The method of least squares or least square regression is to minimise 'S' i.e. sum of the squares of the errors.

Fitting a straight Line (Linear regression)

$$\text{Let, } y = f(x) = a_0 + a_1 x$$

$$\therefore S = [y_1 - (a_0 + a_1 x_1)]^2 + [y_2 - (a_0 + a_1 x_2)]^2 + \dots + [y_n - (a_0 + a_1 x_n)]^2$$

$$\text{i.e. } S = \phi(a_0, a_1)$$

For 'S' to be minimum, we have $\frac{\partial S}{\partial a_0} = 0, \frac{\partial S}{\partial a_1} = 0$.

$$\frac{\partial S}{\partial a_0} = 0 = -2[y_1 - (a_0 + a_1 x_1)] - 2[y_2 - (a_0 + a_1 x_2)] - \dots - 2[y_n - (a_0 + a_1 x_n)]$$

$$\Rightarrow n a_0 + a_1 \sum_{i=1}^n x_i = \sum_{i=1}^n y_i \quad \text{--- (1)}$$

$$\text{For } \frac{\partial S}{\partial a_1} = 0 = -2x_1[y_1 - (a_0 + a_1 x_1)] - 2x_2[y_2 - (a_0 + a_1 x_2)] - \dots - 2x_n[y_n - (a_0 + a_1 x_n)]$$

$$\Rightarrow a_0 \sum_{i=1}^n x_i + a_1 \sum_{i=1}^n x_i^2 = \sum_{i=1}^n x_i y_i \quad \text{--- (2)}$$

Equations ① & ② are called as normal equations, which ~~(21)~~ (33) can be solved for two unknowns a_0 & a_1 .

Again it is evident that $\frac{\partial^2 S}{\partial a_0^2}$ & $\frac{\partial^2 S}{\partial a_1^2}$ are both +ve.
Hence, for a_0 & a_1 S is minimum.

Prob. 22

Find the best fitting straight line by using the following experimental data points.

$(0.5, 0.31), (1.0, 0.82), (1.5, 1.29), (2.0, 1.85), (2.5, 2.51)$ &

If another point $x=3.0, y=3.02$ is considered, find the effect.

Solⁿ -

POINT	0	1	2	3	4	$\Sigma n = 5$
x	0.5	1.0	1.5	2.00	2.50	$\Sigma x_i = 7.5$
y	0.31	0.82	1.29	1.85	2.51	$\Sigma y_i = 6.78$
xy	0.155	0.82	1.935	3.70	6.275	$\Sigma x_i y_i = 12.885$
x^2	0.25	1.0	2.250	4.0	6.25	$\Sigma x_i^2 = 13.75$