

The Gauss-seidel method is given by

$$x_1^{(n+1)} = \frac{1}{2}[7 + x_2^n]$$

$$x_2^{(n+1)} = \frac{1}{2}[1 + x_1^{(n+1)} + x_3^{(n)}]$$

$$x_3^{(n+1)} = \frac{1}{2}[1 + x_2^{(n+1)}]$$

Given, $X^{(0)} = [x_1^{(0)}, x_2^{(0)}, x_3^{(0)}] = [0, 0, 0]$

$$x_1^{(1)} = \frac{1}{2}[7 + x_2^{(0)}] = \frac{1}{2}[7 + 0] = 3.5$$

$$x_2^{(1)} = \frac{1}{2}[1 + x_1^{(1)} + x_3^{(0)}] = \frac{1}{2}[1 + 3.5 + 0] = 2.25$$

$$x_3^{(1)} = \frac{1}{2}[1 + x_2^{(1)}] = \frac{1}{2}[1 + 2.25] = 1.625$$

Hence option (b) is correct.

25. Consider the system of equations

[D.U. 2017]

$$\begin{bmatrix} 1 & -a \\ -a & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

where 'a' is a real constant. Then Gauss-seidal method for the solution of the above system converges for

- (a) all values of a (b) $|a| < 1$ (c) $|a| > 1$ (d) $a > 2$

Soln. Consider the system $AX = B$. We know that the gauss seidal method converges for $AX = B$ if the matrix A is diagonally dominated.

i.e. $a_{ii} > \sum_{i \neq j} |a_{ij}|$ for each i

∴ we have $1 > |a|$

Hence option (b) is correct.

26. The total number of arithmetic operations required to find the solution of a system of n linear equation in n unknowns by Gauss elimination method is [D.U. 2018]

(a) $\frac{2}{3}n^3 + \frac{1}{2}n^2 - \frac{5}{6}n$

(b) $n^3 - \frac{1}{6}n$

(c) $\frac{2}{3}n^3 + \frac{3}{2}n^2 - \frac{7}{6}n$

(d) $\frac{1}{3}n^3 + \frac{1}{2}n^2 - \frac{5}{6}n$

Soln. Article - Counting operation in Gaussian elimination.

Consider a system of n linear equation in n unknowns and the corresponding $n \times n$ coefficient matrix

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ | & | & & | \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$

The $n \times 1$, solution matrix $X = [x_1, x_2, \dots, x_n]^T$, and the $n \times 1$ constant matrix $B = [b_1, b_2, \dots, b_n]^T$.

consider $AX = B$

Then After Gaussian elimination, the system convert as

$$UX = G$$

Where $u = \begin{bmatrix} u_{11} & u_{12} & \dots & u_{1n} \\ & u_{22} & \dots & u_{2n} \\ & & & | \\ & & & \dots & u_{nn} \end{bmatrix}$ is the coefficient matrix of the equivalent system,

and $g = \begin{bmatrix} g_1 \\ g_2 \\ | \\ g_n \end{bmatrix}$ be the $n \times 1$ equivalent constant matrix.

The following table states the operation count from going from A to U at each step as

Step Number	Additional subtraction	Number of multiplications	Divisions
1.	$(n-1)^2$	$(n-1)^2$	$n-1$
2.	$(n-2)^2$	$(n-2)^2$	$n-2$
⋮	⋮	⋮	⋮
$n-2$	4	4	2
$n-1$	1	1	1

Therefore

$$\text{Total number of Addition/Subtraction} = \sum (n-1)^2 = \frac{n(n-1)(2n-1)}{6}$$

$$\text{Total number of Multiplication/Divisions} = \sum (n-1)^2 + \sum (n-1) = \frac{n(n-1)(2n-1)}{6} + \frac{n(n-1)}{2} = \frac{n(n^2-1)}{3}$$

Now we count the Number of Addition/Subtraction and the Number of Multiplication/divisions in going from $b \rightarrow g$, we have

$$\text{Total Number of Addition/Subtraction} = \sum (n-1) = \frac{n(n-1)}{2}$$

$$\text{Total Number of Multiplication/divisions} = \sum (n-1) = \frac{n(n-1)}{2}$$

Lastly, we count the Number of Addition/Subtraction and Multiplications/divisions for finding the solution from the back substitution method, as

$$\text{Total Number of addition/subtraction} = \sum (n-1) = \frac{n(n-1)}{2}$$

$$\text{Total Number of Multiplication/division} = \sum n = \frac{n(n+1)}{2}$$

Therefore, the total number of operations to obtain the solution of a system of n linear equations in n variables using Gaussian - elimination method is

$$\text{Total Number of additions/subtraction} = \frac{n(n-1)(2n-1)}{6} + \frac{n(n-1)}{2} + \frac{n(n-1)}{2} = \frac{n(n-1)(2n+5)}{6}$$

Total number of multiplication /divisions to obtain solution is

$$= \frac{n(n^2 - 1)}{3} + \frac{n(n-1)}{2} + \frac{n(n+1)}{2} = \frac{n(n^2 + 3n + 1)}{3}$$

Therefore total number of Arithmetic operations are

$$= \frac{n(n-1)(2n+5)}{6} + \frac{n(n^2 + 3n - 1)}{3}$$

$$= \frac{4n^3 + 9n^2 - 7n}{6} = \frac{2}{3}n^3 + \frac{3}{2}n^2 - \frac{7}{6}n$$

Total number of operation are

$$= \frac{2}{3}n^3 + \frac{3}{2}n^2 - \frac{7}{6}n$$

Hence option (c) is correct

27. If $\begin{pmatrix} 1 & 4 & 3 \\ 2 & 7 & 9 \\ 5 & 8 & a \end{pmatrix} = \begin{bmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & -53 \end{bmatrix} \begin{bmatrix} 1 & u_{12} & u_{13} \\ 0 & 1 & u_{23} \\ 0 & 0 & 1 \end{bmatrix}$, then the value of a is

[GATE-2008]

- (a) -2 (b) -1 (c) 1 (d) 2

Soln. $\begin{pmatrix} 1 & 4 & 3 \\ 2 & 7 & 9 \\ 5 & 8 & a \end{pmatrix} = \begin{bmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & -53 \end{bmatrix} \begin{bmatrix} 1 & u_{12} & u_{13} \\ 0 & 1 & u_{23} \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} l_{11} & l_{11}u_{12} & l_{11}u_{13} \\ l_{21} & l_{11}u_{12} + l_{22} & l_{21}u_{13} + l_{22}u_{23} \\ l_{31} & l_{31}u_{12} + l_{32} & l_{31}u_{13} + l_{32}u_{23} - 53 \end{bmatrix}$

Comparing both sides, we have

$l_{11} = 1, l_{21} = 2, l_{31} = 5, l_{11}u_{12} = 4$ implies $u_{12} = 4$

$l_{11}u_{13} = 3$ implies $u_{13} = 3, l_{21}u_{12} + l_{22} = 7$ implies $8 + l_{22} = 7$ i.e. $l_{22} = -1$

$l_{21}u_{13} + l_{22}u_{23} = 9$ implies $6 - u_{23} = 9$ i.e. $u_{23} = -3$

$l_{31}u_{12} + l_{32} = 8$ implies $20 + l_{32} = 8$ i.e. $l_{32} = -12$

Now, $l_{31}u_{13} + l_{32}u_{23} - 53 = a$ implies $15 + 36 - 53 = a$ i.e. $a = -2$

Option (a) is correct

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Interpolation by Polynomials

Suppose that a function $f(x)$ is not defined explicitly, but its value at some finite number of points $\{x_i, i=1, 2, \dots, n\}$ is given. The interest is to find the value of f at some point x lying between x_j and x_k , for some $j, k=1, 2, \dots, n$. This can be obtained by first approximating f by a known function and then finding the value of this approximate function at the point x . Such a process is called the interpolation.

Lagrange Interpolation:

The basic interpolation problem can be posed in one of two ways:

I. Given a set of nodes $\{x_i / i=0, 1, \dots, n\}$ and corresponding data values $\{y_i / i=0, 1, \dots, n\}$. Find the polynomial $p_n(x)$ of degree less than or equal to n , such that $p_n(x_i) = y_i, i=0, 1, \dots, n$

II. Given a set of nodes $\{x_i / i=0, 1, \dots, n\}$ and a continuous function $f(x)$. Find the polynomial $p_n(x)$ of degree less than or equal to n , such that $p_n(x_i) = f(x_i), i=0, 1, \dots, n$

Note that in the first problem we are trying to fit a polynomial to the data, and in the second case, we are trying to approximate a given function with the interpolating polynomial. Note that the first problem can be viewed as a particular case of the second.

Theorem (Lagrange Interpolation Formula)

Let $x_0, x_1, \dots, x_n \in I = [a, b]$ be $n+1$ distinct nodes and let $f(x)$ be a continuous real-valued function defined on I . Then, there exists a unique polynomial p_n of degree $\leq n$ (called Lagrange Formula for interpolating Polynomial), given by

$$p_n(x) = \sum_{k=0}^n f(x_k) l_k(x) \text{ where } l_k(x) = \prod_{i=0, i \neq k}^n \frac{x - x_i}{x_k - x_i}, k=0, \dots, n$$

such that $p_n(x_i) = f(x_i), i=0, 1, \dots, n$

The function $l_k(x)$ is called the **Lagrange Multiplier**.

Newton Interpolation and Divided Differences

We have seen that in the Lagrange formula of interpolating polynomial for a function, if we decide to add a point to the set of nodes to increase the accuracy, we have to completely recompute all of the $l_i(x)$ functions. In other words, we cannot express P_{n+1} in terms of p_n , using Lagrange formula. An alternate form of the polynomial, known as the Newton form, avoids this problem, and allows us to easily write P_{n+1} in terms of P_n .

The idea behind the Newton formula of the interpolating polynomial is to write $P_n(x)$ in the form (called Newton form)