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1 Introduction

This method is a variant of **Guass- siedel law** for solving a linear system of equation by iterative method, resulting in faster convergence. The main purpose of this method is to solve the linear systems automatically on the digital systems. These methods were designed for computation by **human calculators**.

This method was introduced by **David M. Young Jr.** and by **Stanley P. Frankel** in 1950

2 Mathematic Form

Suppose, we have n numbers of linear equations with x as unknown. So , it can also be written as:

$$AX = b \tag{1}$$

where,

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix},$$

$$X = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ \vdots \\ x_n \end{pmatrix} \text{ and } b = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ \vdots \\ \vdots \\ b_n \end{pmatrix}$$

Now breaking matrix A in D (diagonal component), L (Lower triangular matrix) and U (upper triangular matrix)

Therefore, A can be written as-

$$A = D + L + U \tag{2}$$

where,

$$A = D + L + U$$

$$D = \begin{bmatrix} a_{11} & 0 & \cdots & 0 \\ 0 & a_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_{nn} \end{bmatrix}, \quad L = \begin{bmatrix} 0 & 0 & \cdots & 0 \\ a_{21} & 0 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & 0 \end{bmatrix}$$

$$U = \begin{bmatrix} 0 & a_{12} & \cdots & a_{1n} \\ 0 & 0 & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 \end{bmatrix}.$$
tion (1) can be written as -

and

$$U = \begin{bmatrix} 0 & a_{12} & \cdots & a_{1n} \\ 0 & 0 & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 \end{bmatrix}$$

Now, equation (1) can be written as -

$$(D + \omega L)\mathbf{x} = \omega \mathbf{b} - [\omega U + (\omega - 1)D]\mathbf{x}$$
(3)

constant ω is called the the relaxation factor. As, we already mentioned that it is a iterative method so we can solve the left hand side of the above equation for x using the previous value of x from the right hand side.

Now using **iteration** it can be written as:

$$\mathbf{x}^{(k+1)} = (D + \omega L)^{-1} (\omega \mathbf{b} - [\omega U + (\omega - 1)D]\mathbf{x}^{(k)}) = L_w \mathbf{x}^{(k)} + \mathbf{c}$$

$$X^k$$
 is the n^{th} approximation. X^{k+1} is the next approximation to n^{th} .

Now, by using **forward substitution** it can be written as:

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

where,
$$i = 1, 2, \dots, n$$
.