

LU decomposition:

A ~~n x n~~ $n \times m$ matrix which is reduced to a row-echelon form by Gauss-elimination method used certain square of elementary row operations.

- A factorization of a square matrix A as $A = LU$, where L is lower triangular and U is the upper triangular matrix is called an LU decomposition (or) triangular decomposition of A

Applications:

1. A solution of a system of equations, which itself is an integral part of many applications such as ~~finding~~ finding current in a circuit
 - finding current in a circuit
 - solution of discrete dynamical system problems
2. finding inverse of a matrix
3. finding determinant of the matrix

Advantages:

1. Once A is decomposed, we can solve $AX=b$ for as many constant vector b as we please [different values of b for the same A]
2. The cost of each additional solution is relatively small, because the forward and backward substitution operations are much less time consuming than the decomposition process

Disadvantages:

1. It requires forward and backward substitution
2. Solving requires storing in memory the LU factors

Doolittle's decomposition

Decomposition phase: Doolittle's decomposition is closely related to Gauss elimination.

1. Create matrices: A , X and B , where A is the augmented matrix, X constitutes the variable vectors and B are the constants.

2. Let $A = LU$, where L is the lower triangular and U is the upper triangular ~~matrix~~;

Diagonal terms in the lower triangle should be equal one

$$A = LU$$

$$A = \begin{bmatrix} 1 & 0 & 0 \\ a & 1 & 0 \\ b & c & 1 \end{bmatrix} \begin{bmatrix} d & e & f \\ 0 & g & h \\ 0 & 0 & i \end{bmatrix}$$

Gauss Jacobi Iteration Method:

This method is applicable to ~~determin~~ the system of equation in which leading diagonal elements of co-efficient matrix are dominant (large in magnitude) in their respective rows.

Working rule :- Consider the system of eqns

$$a_{11}x + a_{12}y + a_{13}z = b_1$$

$$a_{21}x + a_{22}y + a_{23}z = b_2$$

$$a_{31}x + a_{32}y + a_{33}z = b_3$$

Note :- Diagonal dominance property must be satisfied

$$\text{i.e. } |a_{11}| > |a_{12}| + |a_{13}|$$

$$|a_{22}| > |a_{21}| + |a_{23}|$$

$$|a_{33}| > |a_{31}| + |a_{32}|$$

- Rewriting the eqns for x, y, z respectively

$$x = \frac{1}{a_{11}} (b_1 - a_{12}y - a_{13}z)$$

$$y = \frac{1}{a_{22}} (b_2 - a_{21}x - a_{23}z)$$

$$z = \frac{1}{a_{33}} (b_3 - a_{31}x - a_{32}y)$$

Iteration 1: Put $x = x_0, y = y_0, z = z_0$

$$x_1 = \frac{1}{a_{11}} (b_1 - a_{12}y_0 - a_{13}z_0)$$

$$y_1 = \frac{1}{a_{22}} (b_2 - a_{21}x_0 - a_{23}z_0)$$

$$z_1 = \frac{1}{a_{33}} (b_3 - a_{31}x_0 - a_{32}y_0)$$

The above iteration process is continued until two successive approximations are equal.

Advantages:

1. It can be done in parallel
2. It is highly desirable for many applications

Disadvantages:

1. It is very lengthy and time taking
2. Inflexible

Application

Jacobi method is an iterative algorithm for determining the solutions of a strictly diagonally dominant system of linear equations.

Convergence:

It converges if A is strictly diagonally dominant.

Jacobi method in matrix form

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

$A \quad X \quad B$

$$L = \begin{bmatrix} 0 & 0 & 0 \\ a_{21} & 0 & 0 \\ a_{31} & a_{32} & 0 \end{bmatrix}, \quad D = \begin{bmatrix} a_{11} & 0 & 0 \\ 0 & a_{22} & 0 \\ 0 & 0 & a_{33} \end{bmatrix}$$

$$U = \begin{bmatrix} 0 & a_{12} & a_{13} \\ 0 & 0 & a_{23} \\ 0 & 0 & 0 \end{bmatrix}$$

$$A = (L + D + U)$$

$$X^{k+1} = \frac{1}{D} [B - (L + U)X^k]$$

Stopping criteria :-

$$\|x^k - x^{k+1}\| \leq \epsilon, \quad \epsilon > 0$$

$$x^k \approx x^{k+1}$$

Gauss - Seidel Method

This method is also known as the Liebmann method or the method of successive displacement. It is an iterative method to solve a system of linear equations.

- Here immediate updation will be taken place.

Working rule:

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 = b_1$$

$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 = b_2$$

$$a_{31}x_1 + a_{32}x_2 + a_{33}x_3 = b_3$$

Note: Diagonal dominance property must be satisfied

$$|a_{11}| > |a_{12}| + |a_{13}|$$

$$|a_{22}| > |a_{21}| + |a_{23}|$$

$$|a_{33}| > |a_{31}| + |a_{32}|$$

$$x_1 = \frac{1}{a_{11}} (b_1 - (a_{12}x_2 + a_{13}x_3))$$

$$x_2 = \frac{1}{a_{22}} (b_2 - (a_{21}x_1 + a_{23}x_3))$$

$$x_3 = \frac{1}{a_{33}} (b_3 - (a_{31}x_1 + a_{32}x_2))$$

Initial approximation

Ist iteration $x_1^{(0)} = 0, x_2^{(0)} = 0, x_3^{(0)} = 0$

$$x_1' = \frac{1}{a_{11}} (b_1 - (a_{12}x_2^{(0)} + a_{13}x_3^{(0)}))$$

$$x_2' = \frac{1}{a_{22}} (b_2 - (a_{21}x_1^{(1)} + a_{23}x_3^{(0)}))$$

$$x_3' = \frac{1}{a_{33}} (b_3 - (a_{31}x_1^{(1)} + a_{32}x_2^{(1)}))$$

IInd iteration

$$x_1^{(2)} = \frac{1}{a_{11}} (b_1 - (a_{12}x_2^{(1)} + a_{13}x_3^{(1)}))$$

$$x_2^{(2)} = \frac{1}{a_{22}} (b_2 - (a_{21}x_1^{(2)} + a_{23}x_3^{(1)}))$$

$$x_3^{(2)} = \frac{1}{a_{33}} (b_3 - (a_{31}x_1^{(2)} + a_{32}x_2^{(2)}))$$

The above iteration process is continued until two successive approximations are equal

Advantages :-

- 1) It requires less iteration
- 2) Relatively easy to program

Disadvantages :-

- 1) Not suitable for large systems
- 2) Required more no. of iterations to reach convergence
- 3) Convergence time increases with size of the system

Applications :-

1. It is applicable to strictly diagonally dominant or symmetric ^{positive} definite matrices because only in this case convergence is possible
- 2.

Convergence :-

The soln of linear eqⁿ by iterative methods requires for convergence that the absolute magnitude of all the eigenvalues of the iteration matrix should be less than unity

2. G-S method converges if the no. of roots inside the unit circle is equal to the order of the iteration matrix

General form:-

$$x_1^{k+1} = \frac{1}{a_{11}} [b_1 - (a_{12}x_2^k + a_{13}x_3^k)]$$

$$x_2^{k+1} = \frac{1}{a_{22}} [b_2 - (a_{21}x_1^{k+1} + a_{23}x_3^k)]$$

$$x_3^{k+1} = \frac{1}{a_{33}} [b_3 - (a_{31}x_1^{k+1} + a_{32}x_2^{k+1})]$$

\downarrow
D

B L + U

$$X^{k+1} = \frac{1}{D} [B - (LX^{k+1} + UX^k)]$$

$$\|X^{k+1} - X^k\| \leq 0.001$$

$$X^{k+1} = (D+L)^{-1}B - (D+L)^{-1}UX^k$$