

## **Description of Gauss Elimination Method:**

Gauss elimination can be divided into three parts

1. Partial pivoting:

Partial pivoting is a crucial step to avoid division by zero. The diagonal elements are checked for zero of a row and replaced by non zero element by row interchange operation.

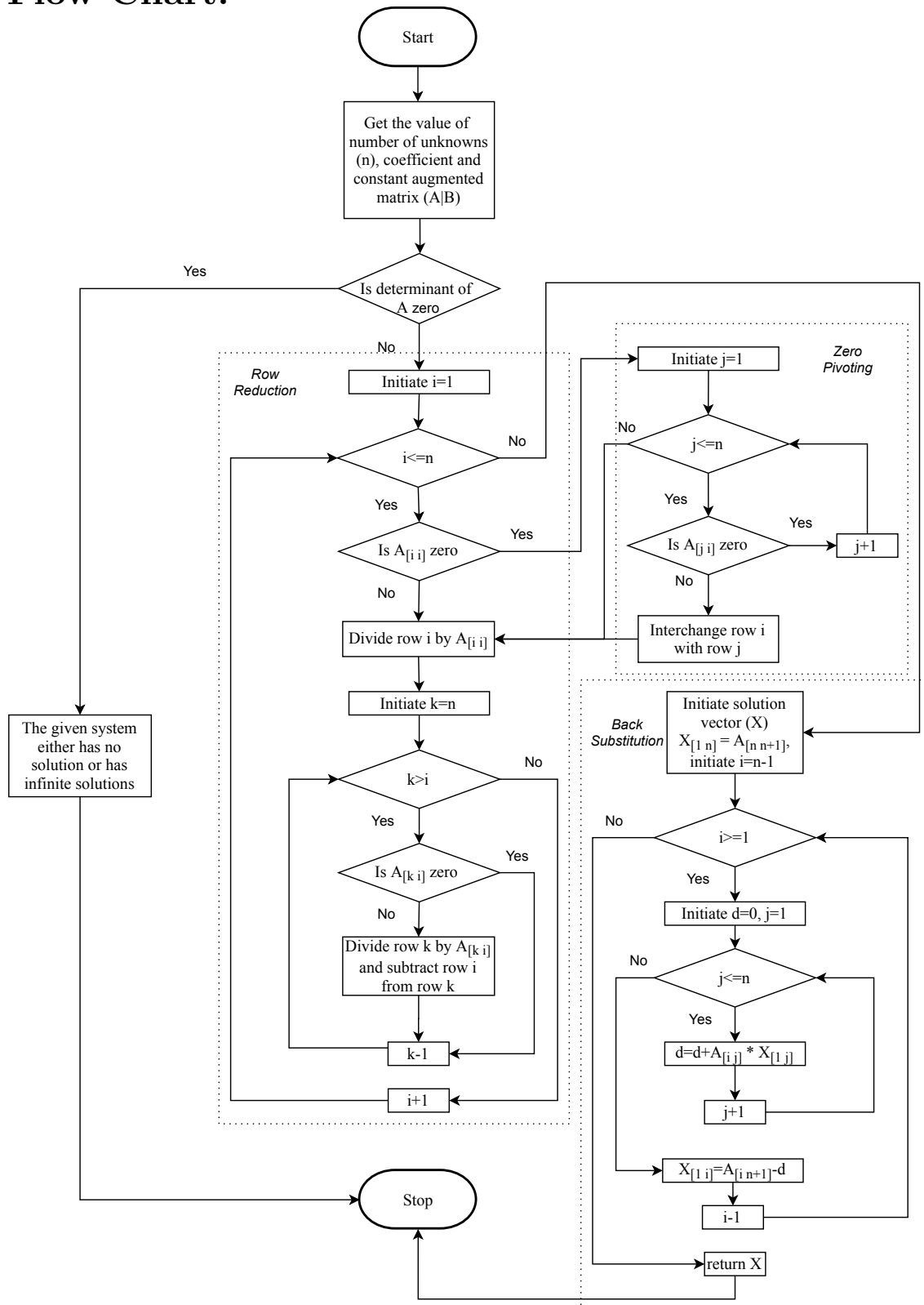
2. Row reduction:

The coefficient matrix is transformed into an upper triangular matrix by row operations. A simple approach is to divide a row by it's diagonal element, divide the succeeding row by element corresponding to the diagonal element of the former row and subtract the former row from the later row. If the succeeding row has zero at the place corresponding to the diagonal element, skip the row. Repeating this for all the rows and their diagonal elements will give an upper triangular matrix.

3. Back substitution:

The coefficient matrix thus can be used to find the unknowns in reverse order by substituting the values.

## Flow Chart:



## Description of Jacobi's Method:

The Jacobi's method is an iterative technique for solving a system of  $n$  linear equations with unknown  $X$ :

$$AX=B$$

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_{11} \\ x_{21} \\ \vdots \\ x_{n1} \end{bmatrix} = \begin{bmatrix} b_{11} \\ b_{21} \\ \vdots \\ b_{n1} \end{bmatrix}$$

The solution is then obtained by iteration via

$$x_{in}^{k+1} = \frac{1}{a_{ii}} \left( b_{in} - \sum_{j \neq i} a_{ij} x_{jn}^k \right), \quad i = 1, 2, \dots, n.$$

where  $X^k$  is the  $k$ th approximation of  $X$  and  $X^{k+1}$  is the next or  $k+1$  iteration of  $X$ . The computation of  $x_i^{k+1}$  requires each element in  $x^k$  except itself. Unlike Gauss Seidel method, we can't overwrite  $x_i^k$  with  $x_i^{k+1}$ , as the value will be needed by the rest of the computation. The minimum amount of storage is two vectors of size  $n$ .

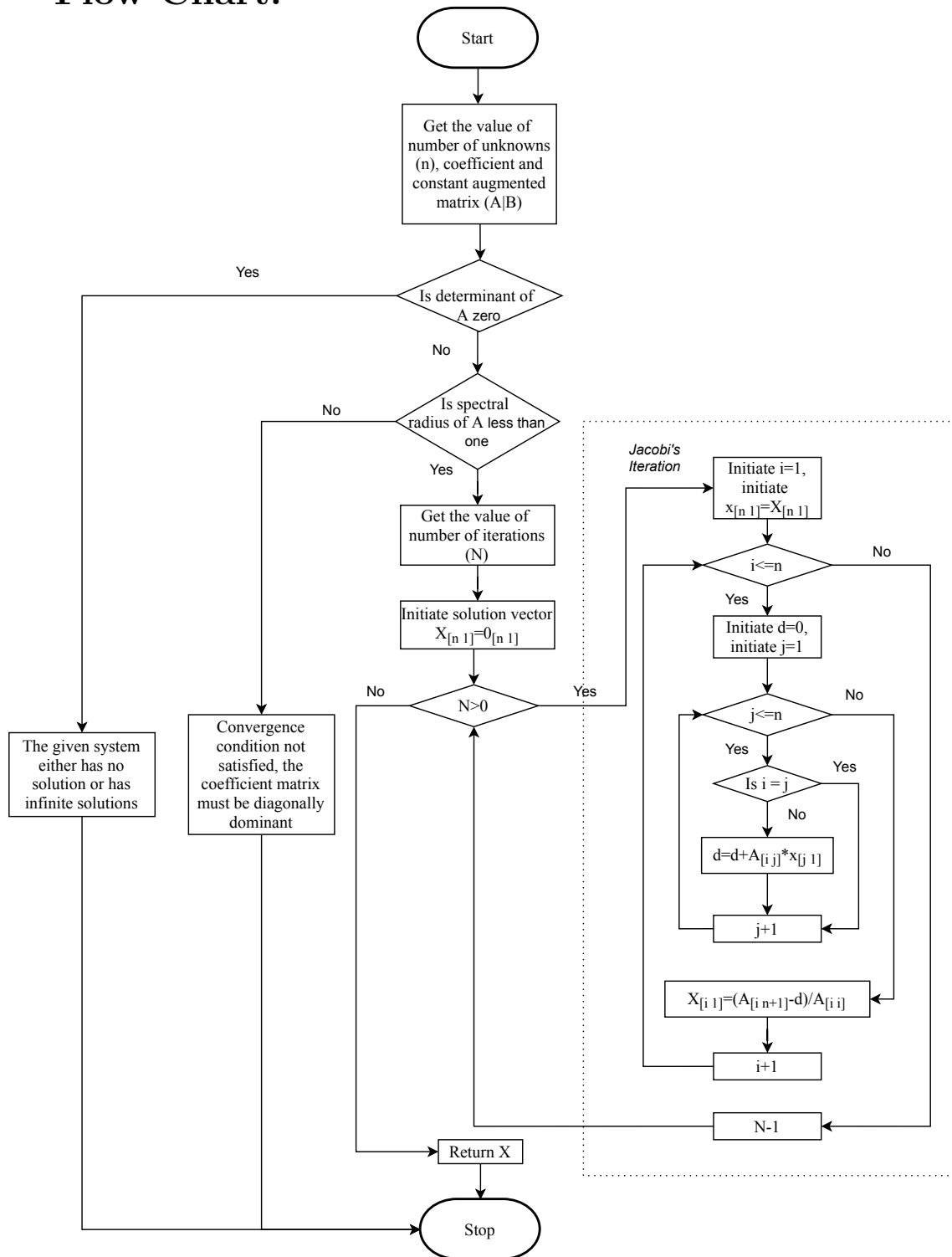
The standard convergence condition for any iterative method is the spectral radius of the iteration matrix should be less than 1:

$$\rho(D^{-1}(L + U)) < 1$$

where  $A$  is decomposed into its diagonal component  $D$ , a lower triangular part  $L$  and an upper triangular part  $U$ .  $D^{-1}$  is inverse of  $D$ .

The spectral radius of a square matrix is the largest absolute value of its eigenvalues. It is denoted by  $\rho(\cdot)$ .

## Flow Chart:



## Description of Gauss Seidel Method:

The Gauss Seidel method is an iterative technique for solving a system of  $n$  linear equations with unknown  $X$ :

$$AX=B$$

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_{11} \\ x_{21} \\ \vdots \\ x_{n1} \end{bmatrix} = \begin{bmatrix} b_{11} \\ b_{21} \\ \vdots \\ b_{n1} \end{bmatrix}$$

The solution is then obtained by iteration via

$$x_{in}^{k+1} = \frac{1}{a_{ii}} \left( b_{in} - \sum_{j \neq i} a_{ij} x_{jn}^k \right), \quad i = 1, 2, \dots, n.$$

where  $X^k$  is the  $k$ th approximation of  $X$  and  $X^{k+1}$  is the next or  $k + 1$  iteration of  $X$ . The computation of  $X^{k+1}$  uses the elements of  $X^{k+1}$  that have already been computed and elements of  $x^k$  that have not been computed in the  $k + 1$  iteration. Thus unlike the Jacobi's method, only one storage vector is required as elements can be overwritten as they are computed.

The standard convergence condition for any iterative method is the spectral radius of the iteration matrix should be less than 1:

$$\rho(D^{-1}(L + U)) < 1$$

where  $A$  is decomposed into its diagonal component  $D$ , a lower triangular part  $L$  and an upper triangular part  $U$ .  $D^{-1}$  is inverse of  $D$ .

The spectral radius of a square matrix is the largest absolute value of its eigenvalues. It is denoted by  $\rho(\cdot)$ .

## Flow Chart:

