~ Linear Algebraic Equations ~

# **Gauss Elimination**

**Chapter 9** 

# Solving Systems of Equations

• A linear equation in n variables:

$$a_1x_1 + a_2x_2 + \dots + a_nx_n = b$$

- For small ( $n \le 3$ ), linear algebra provides several tools to solve such systems of linear equations:
  - Graphical method
  - Cramer's rule
  - Method of elimination
- Nowadays, easy access to computers makes the solution of **very large** sets of linear algebraic equations possible

# Determinants and Cramer's Rule

$$[A]\{x\} = \{b\}$$

[A]: coefficient matrix

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

$$x_{1} = \frac{\begin{vmatrix} b_{1} & a_{12} & a_{13} \\ b_{2} & a_{22} & a_{23} \\ b_{3} & a_{32} & a_{33} \end{vmatrix}}{D}$$

$$x_{2} = \frac{\begin{vmatrix} a_{11} b_{1} a_{13} \\ a_{21} b_{2} a_{23} \\ a_{31} b_{3} a_{33} \end{vmatrix}}{D}$$

$$x_{3} = \frac{\begin{vmatrix} a_{11} & a_{12} & b_{1} \\ a_{21} & a_{22} & b_{2} \\ a_{31} & a_{32} & b_{3} \end{vmatrix}}{D}$$

D: Determinant of A matrix

# **Computing the Determinant**

$$D = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$$

$$[A]\{x\} = \{B\}$$

$$D_{11} = \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} = a_{22} a_{33} - a_{32} a_{23}$$

$$[A] = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$D_{12} = \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} = a_{21} a_{33} - a_{31} a_{23}$$

$$D_{13} = \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix} = a_{21} a_{32} - a_{31} a_{22}$$

$$a_{31} a_{32} = a_{21} a_{32} - a_{31} a_{22}$$

Determinant of A = D = 
$$a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

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# Gauss Elimination

- Solve Ax = b
- Consists of two phases:
  - -Forward elimination
  - -Back substitution
- Forward Elimination reduces Ax = b to an upper triangular system Tx = b'
- *Back substitution* can then solve Tx = b' for x

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

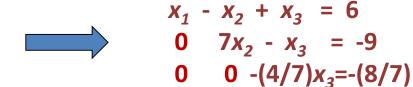
Forward Elimination

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & b_1 \\ 0 & a_{22} & a_{23} & b_2 \\ 0 & 0 & a_{33} & b_3 \end{bmatrix}$$

Back Substitution

$$x_{3} = \frac{b_{3}^{"}}{a_{33}^{"}} \quad x_{2} = \frac{b_{2}^{"} - a_{23}^{"} x_{3}}{a_{22}^{"}}$$

$$x_{1} = \frac{b_{1} - a_{13} x_{3} - a_{12} x_{2}}{a_{11}}$$



\*\*here Solve using BACK SUBSTITUTION: 
$$x_3 = 2$$
  $x_2 = -1$   $x_1 = 3$ 

$$1x_0 + 1x_1 - 1x_2 + 4x_3 = 8$$

$$-2x_1$$
  $-3x_2$   $+1x_3$  = 5

$$2x_2 - 3x_3 = 0$$

$$x_3 = 2$$
  $2x_3 = 4$ 

$$1x_0 + 1x_1 - 1x_2$$

$$-2x_1$$
  $-3x_2$ 

$$x_2 = 3$$

$$2x_2$$

6

$$1x_0 + 1x_1$$

$$x_1 = -6 \qquad -2x_1$$

 $x_0 = 9$ 

 $1x_0$ 

=

9

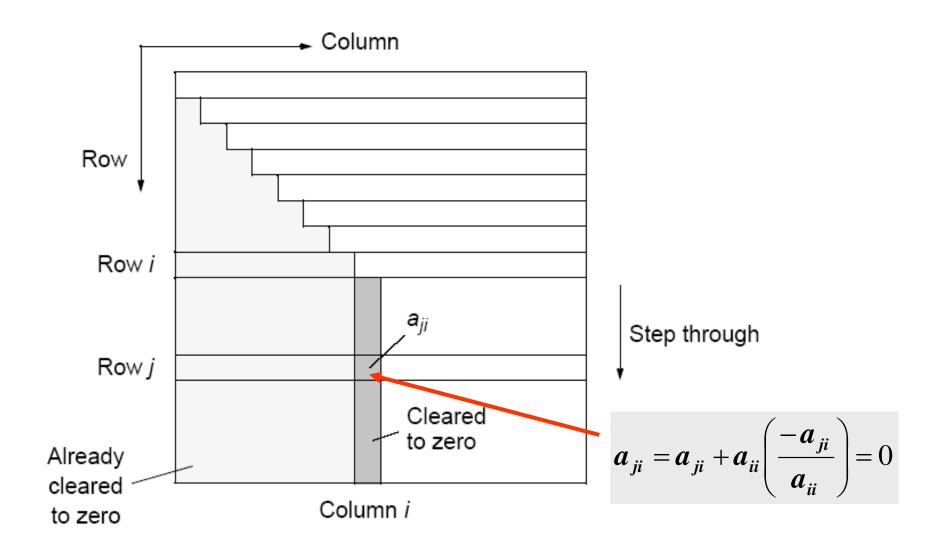
(\* Pseudocode \*)

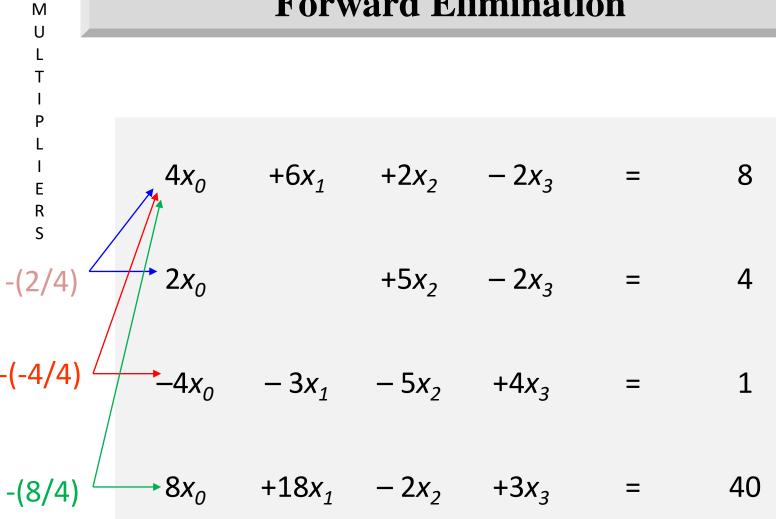
```
for i \leftarrow n down to 1 do
         /* calculate x_i */
         x[i] \leftarrow b[i]/a[i,i]
         /* substitute x [ i ] in the equations above */
         for j \leftarrow 1 to i-1 do
                  b[i] \leftarrow b[i] - x[i] \times a[i,i]
         endfor
endfor
```

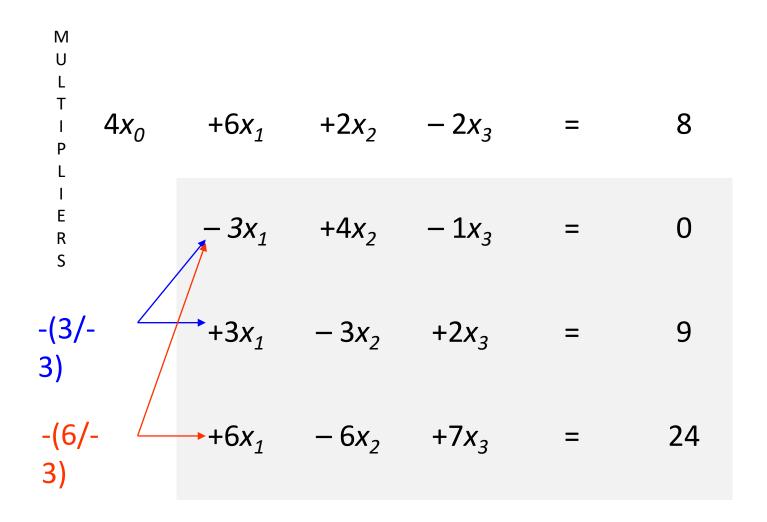
**Time Complexity?** 

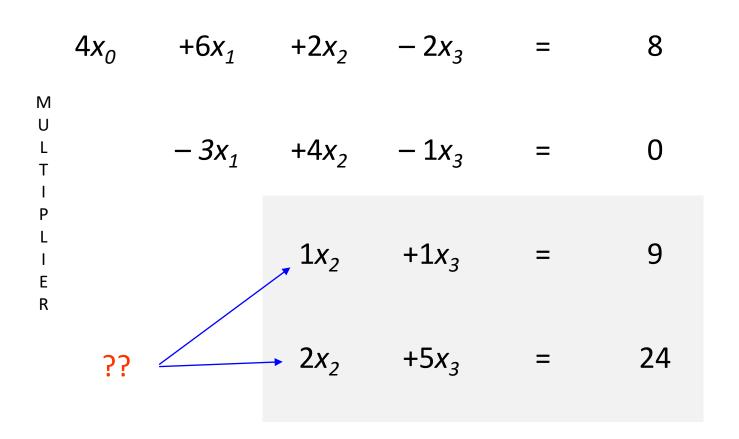


 $O(n^2)$ 









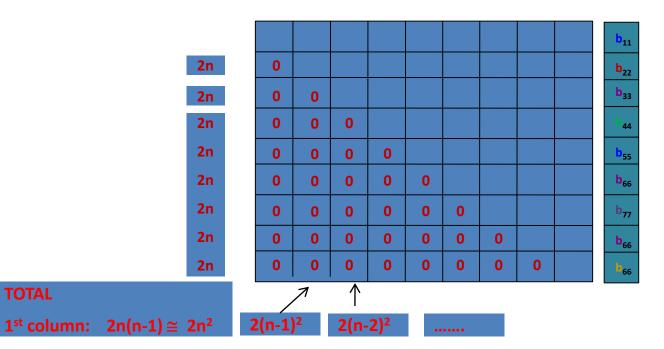
$$4x_0 +6x_1 +2x_2 -2x_3 = 8$$

$$-3x_1 +4x_2 -1x_3 = 0$$

$$1x_2 +1x_3 = 9$$

$$3x_3 = 6$$

# **Operation count in Forward Elimination**



**TOTAL** 

17

TOTAL# of Operations for FORWARDELIMINATION:

$$2n^{2} + 2(n-1)^{2} + \dots + 2*(2)^{2} + 2*(1)^{2} = 2\sum_{i=1}^{n} i^{2}$$

$$= 2\frac{n(n+1)(2n+1)}{6}$$

$$= O(n^{3})$$

## **Pitfalls of Elimination Methods**

## Division by zero

It is possible that during both elimination and back-substitution phases a division by zero can occur.

### For example:

$$2x_2 + 3x_3 = 8$$
 0 2 3  
 $4x_1 + 6x_2 + 7x_3 = -3$  A = 4 6 7  
 $2x_1 + x_2 + 6x_3 = 5$  2 1 6

Solution: *pivoting* (to be discussed later)

# Pitfalls (cont.)

## Round-off errors

- Because computers carry only a limited number of significant figures, round-off errors will occur and they will *propagate* from one iteration to the next.
- This problem is especially important when **large** numbers of equations (100 or more) are to be solved.
- Always use **double-precision** numbers/arithmetic. It is slow but needed for correctness!
- It is also a good idea to substitute your results back into the original equations and check whether a substantial error has occurred.

# Pitfalls (cont.)

ill-conditioned systems - small changes in coefficients result in large changes in the solution. Alternatively, a wide range of answers can approximately satisfy the equations.

(Well-conditioned systems – small changes in coefficients result in small changes in the solution)

**Problem:** Since round off errors can induce small changes in the coefficients, these changes can lead to large solution errors in *ill-conditioned* systems.

### **Example:**

$$x_1 + 2x_2 = 10$$

$$\mathbf{1.1}x_1 + 2x_2 = 10.4$$

Example:  

$$x_1 + 2x_2 = 10$$
  
 $1x_1 + 2x_2 = 10.4$   
 $x_1 = \frac{\begin{vmatrix} b_1 \ a_{12} \end{vmatrix}}{D} = \frac{\begin{vmatrix} 10 \ 2 \end{vmatrix}}{1(2) - 2(1.1)} = \frac{2(10) - 2(10.4)}{-0.2} = 4$   $x_2 = 3$ 

$$x_{1} + 2x_{2} = 10$$

$$1.05x_{1} + 2x_{2} = 10.4$$

$$x_{1} = \frac{\begin{vmatrix} b_{1} \ a_{12} \\ b_{2} \ a_{22} \end{vmatrix}}{D} = \frac{\begin{vmatrix} 10 & 2 \\ 10.4 & 2 \end{vmatrix}}{1(2) - 2(1.05)} = \frac{2(10) - 2(10.4)}{-0.1} = 8$$

$$x_{2} = 1$$

# ill-conditioned systems (cont.) -

• Surprisingly, **substitution** of the <u>erroneous</u> values,  $x_1$ =8 and  $x_2$ =1, into the original equation **will not** reveal their incorrect nature clearly:

$$x_1 + 2x_2 = 10$$
 8+2(1) = 10 (the same!)  
1.1 $x_1 + 2x_2 = 10.4$  1.1(8)+2(1)=10.8 (close!)

#### **IMPORTANT OBSERVATION:**

An ill-conditioned system is one with a *determinant* close to **zero** 

- If determinant D=0 then there are infinitely many solutions  $\rightarrow$  *singular system*
- **Scaling** (multiplying the coefficients with the same value) does not change the equations but changes the value of the determinant in a significant way. **However, it does not change the** *ill-conditioned* **state of the equations!**

DANGER! It may hide the fact that the system is ill-conditioned!!

### How can we find out whether a system is ill-conditioned or not?

Not easy! Luckily, most engineering systems yield well-conditioned results!

• One way to find out: change the coefficients slightly and recompute & compare

# Techniques for Improving Solutions

• Use of more significant figures – *double precision arithmetic* 

## • Pivoting

If a pivot element is zero, normalization step leads to division by zero. The same problem may arise, when the pivot element is close to zero. Problem can be avoided:

### - Partial pivoting

Switching the rows below so that the largest element is the pivot element.

\*\*here\* Go over the solution in: CHAP9e-Problem-11.doc

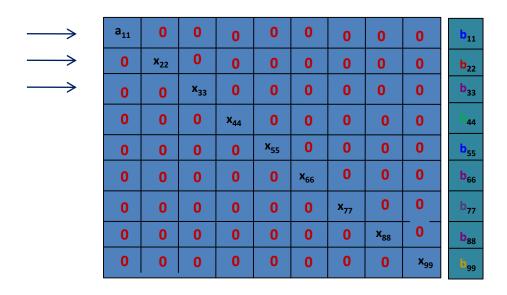
### - Complete pivoting

- Searching for the largest element in all rows and columns then switching.
- This is rarely used because switching columns changes the order of x's and adds significant complexity and overhead → costly

### • Scaling

- used to reduce the round-off errors and improve accuracy

# Gauss-Jordan Elimination



# Gauss-Jordan Elimination: Example

$$\begin{bmatrix} 1 & 1 & 2 \\ -1 & -2 & 3 \\ 3 & 7 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 8 \\ 1 \\ 10 \end{bmatrix}$$

Augmented Matrix : 
$$\begin{bmatrix} 1 & 1 & 2 & 8 \\ -1 & -2 & 3 & 1 \\ 3 & 7 & 4 & 10 \end{bmatrix}$$

R3 ← R3 - (3)R1

$$0 - 1 \quad 5 \mid 9$$

$$0 \quad 4 \quad -2|-14$$

$$R2 \leftarrow R2/(-1)$$

$$0 \quad 1 \quad -5 \mid -9$$

$$0 \ 4 \ -2|-14$$

R3 ← R3-(4)R2

$$\begin{bmatrix} 1 & 0 & 7 & | 17 \end{bmatrix}$$

$$0 \ 1 \ -5|-9$$

#### Scaling R3:

$$R3 \leftarrow R3/(18)$$

$$0 \quad 1 \quad -5| \quad -9$$

$$0 \quad 1 \quad 0|-2.888$$

#### **RESULT:**

$$x_1$$
=8.45,

$$x_2 = -2.89$$
,

$$x_3 = 1.23$$

# Systems of Nonlinear Equations

Locate the roots of a set of simultaneous nonlinear equations:

$$f_1(x_1, x_2, x_3, ..., x_n) = 0$$
  
 $f_2(x_1, x_2, x_3, ..., x_n) = 0$   
:  
:  
:  
:

### Example:

$$x_1^2 + x_1 x_2 = 10$$
  $\Rightarrow$   $f_1(x_1, x_2) = x_1^2 + x_1 x_2 - 10 = 0$   
 $x_2 + 3x_1 x_2^2 = 57$   $\Rightarrow$   $f_2(x_1, x_2) = x_2 + 3x_1 x_2^2 - 57 = 0$ 

• **First-order** Taylor series expansion of a function with more than one variable:

$$\begin{split} f_{1(i+1)} &= f_{1(i)} + \frac{\partial f_{1(i)}}{\partial x_1} (x_{1(i+1)} - x_{1(i)}) + \frac{\partial f_{1(i)}}{\partial x_2} (x_{2(i+1)} - x_{2(i)}) = 0 \\ f_{2(i+1)} &= f_{2(i)} + \frac{\partial f_{2(i)}}{\partial x_1} (x_{1(i+1)} - x_{1(i)}) + \frac{\partial f_{2(i)}}{\partial x_2} (x_{2(i+1)} - x_{2(i)}) = 0 \end{split}$$

• The root of the equation occurs at the value of  $x_1$  and  $x_2$  where  $f_{I(i+1)} = 0$  and  $f_{2(i+1)} = 0$ Rearrange to solve for  $x_{I(i+1)}$  and  $x_{2(i+1)}$ 

$$\begin{split} \frac{\partial f_{1(i)}}{\partial x_1} \, x_{1(i+1)} + \frac{\partial f_{1(i)}}{\partial x_2} \, x_{2(i+1)} &= -f_{1(i)} + \frac{\partial f_{1(i)}}{\partial x_1} \, x_i + \frac{\partial f_{1(i)}}{\partial x_2} \, x_{2(i)} \\ \frac{\partial f_{2(i)}}{\partial x_1} \, x_{1(i+1)} + \frac{\partial f_{2(i)}}{\partial x_2} \, x_{2(i+1)} &= -f_{2(i)} + \frac{\partial f_{2(i)}}{\partial x_1} \, x_i + \frac{\partial f_{2(i)}}{\partial x_2} \, x_{2(i)} \end{split}$$

$$\begin{bmatrix} \frac{\partial f_{1(i)}}{\partial x_1} & \frac{\partial f_{1(i)}}{\partial x_2} \\ \frac{\partial f_{2(i)}}{\partial x_1} & \frac{\partial f_{2(i)}}{\partial x_2} \end{bmatrix} \begin{bmatrix} x_{1(i+1)} \\ x_{2(i+1)} \end{bmatrix} = - \begin{bmatrix} f_{1(i)} \\ f_{2(i)} \end{bmatrix} + \begin{bmatrix} \frac{\partial f_{1(i)}}{\partial x_1} & \frac{\partial f_{1(i)}}{\partial x_2} \\ \frac{\partial f_{2(i)}}{\partial x_1} & \frac{\partial f_{2(i)}}{\partial x_2} \end{bmatrix} \begin{bmatrix} x_{1(i)} \\ x_{2(i)} \end{bmatrix}$$

- Since  $x_{1(i)}$ ,  $x_{2(i)}$ ,  $f_{1(i)}$ , and  $f_{2(i)}$  are all known at the  $i^{\text{th}}$  iteration, this represents a set of two linear equations with two unknowns,  $x_{1(i+1)}$  and  $x_{2(i+1)}$
- You may use several techniques to solve these equations

# General Representation for the solution of Nonlinear Systems of Equations

$$f_{1}(x_{1}, x_{2}, x_{3}, \dots, x_{n}) = 0$$

$$f_{2}(x_{1}, x_{2}, x_{3}, \dots, x_{n}) = 0$$

$$\vdots$$

$$f_{n}(x_{1}, x_{2}, x_{3}, \dots, x_{n}) = 0$$

$$J = \begin{bmatrix} \frac{\partial f_{1}}{\partial x_{1}} & \frac{\partial f_{1}}{\partial x_{2}} & \dots & \frac{\partial f_{1}}{\partial x_{n}} \\ \frac{\partial f_{2}}{\partial x_{1}} & \frac{\partial f_{2}}{\partial x_{2}} & \dots & \frac{\partial f_{2}}{\partial x_{n}} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{\partial f_{n}}{\partial x_{1}} & \frac{\partial f_{n}}{\partial x_{2}} & \dots & \frac{\partial f_{n}}{\partial x_{n}} \end{bmatrix}$$

**←** Jacobian Matrix

Solve this set of linear equations at each iteration:

$$[J_i]{X_{i+1}} = -{F_i} + [J_i]{X_i}$$

$$\begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}_{i=1} = - \begin{bmatrix} f_1(x_1, x_2, x_3, \dots, x_n) \\ f_2(x_1, x_2, x_3, \dots, x_n) \\ \vdots & \vdots & \vdots \\ f_n(x_1, x_2, x_3, \dots, x_n) \end{bmatrix}_{i} + \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}_{i} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}_{i}$$

 $\{X\}_{(i+1)}$  values are found by solving this system of linear equations.

These iterations will continue until the convergence is reached;

which will happen when the approximate relative error (defined on the vector of values) falls below the desired error tolerance E<sub>tol</sub>

# Solution of Nonlinear Systems of Equations

Solve this set of linear equations at each iteration:  $[J_i]\{X_{i+1}\} = -\{F_i\} + [J_i]\{X_i\}$ 

$$[J_i]{X_{i+1}} = -{F_i} + [J_i]{X_i}$$

Rearrange:

$$[J_i]\{X_{i+1} - X_i\} = -\{F_i\}$$
$$[J_i]\{\Delta X_{i+1}\} = -\{F_i\}$$

$$\begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}_i \begin{bmatrix} \Delta x_1 \\ \Delta x_2 \\ \vdots \\ \Delta x_n \end{bmatrix}_{i+1} = - \begin{bmatrix} f_1(x_1, x_2, x_3, \dots, x_n) \\ f_2(x_1, x_2, x_3, \dots, x_n) \\ \vdots \\ f_n(x_1, x_2, x_3, \dots, x_n) \end{bmatrix}_{ii}$$

# Notes on the solution of **Nonlinear Systems of Equations**

*n* non-linear equations

n non-linear equations n unknowns to be solved: 
$$f_1(x_1, x_2, x_3, ..., x_n) = 0$$
 
$$f_2(x_1, x_2, x_3, ..., x_n) = 0$$
 
$$\vdots$$
 
$$f_n(x_1, x_2, x_3, ..., x_n) = 0$$

Jacobian **Matrix:** 

$$J = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \cdots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \cdots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}$$

Need to solve this set of Linear Equations

$$[J_i]{\Delta X_{i+1}} = -{F_i}$$

in each and every iterative solution of the original Non-Linear system of equations.

- For *n* unknowns, the size of the *Jacobian* matrix is  $n^2$  and the time it takes to solve the *linear system of equations* (one iteration of the non-linear system) is proportional to  $O(n^3)$
- There is no guarantee for the convergence of the non-linear system. There may also be slow convergence.
- In summary, finding the solution set for the Non-Linear system of equations is an extremely compute-intensive task.