

Lecture Note-Numerical Analysis (4): Roots of Nonlinear Algebraic Equations

1. Background information on the numerical approximation of the derivative of a function

FDM: Finite Difference Method (유한차분법)을 이용한 도함수 계산

- Taylor series expansion of $f(x+h)$ for a small value h around a given point x

$$f(x+h) \approx f(x) + f'(x)h + \frac{1}{2}f''(x)h^2 + \frac{1}{6}f^{(3)}(x)h^3 + O(h^4)$$

$$f(x-h) \approx f(x) - f'(x)h + \frac{1}{2}f''(x)h^2 - \frac{1}{6}f^{(3)}(x)h^3 + O(h^4)$$

- The 1st order approximation of the derivative of $f(x)$ using one of the above equation

$$f'(x) \approx \frac{1}{h} \{f(x+h) - f(x)\} + \frac{1}{2}f''(x)h + \frac{1}{6}f^{(3)}(x)h^2 + O(h^3) \approx \frac{1}{h} \{f(x+h) - f(x)\} + O(h)$$

or

$$f'(x) \approx \frac{1}{h} \{f(x) - f(x-h)\} + \frac{1}{2}f''(x)h - \frac{1}{6}f^{(3)}(x)h^2 + O(h^3) \approx \frac{1}{h} \{f(x) - f(x-h)\} + O(h)$$

Therefore, the first order numerical approximation becomes

$$\boxed{f'(x) \approx \frac{1}{h} \{f(x+h) - f(x)\}} \quad \text{or} \quad \boxed{f'(x) \approx \frac{1}{h} \{f(x) - f(x-h)\}}$$

,which is called by the forward/backward difference formula

- The 2nd order approximation of the derivative of f(x) by subtracting the above equation

$$f(x+h) - f(x-h) \approx 2f'(x)h + \frac{2}{6}f^{(3)}(x)h^3 + O(h^4)$$

$$f'(x) \approx \frac{1}{2h}\{f(x+h) - f(x-h)\} - \frac{1}{6}f^{(3)}(x)h^2 + O(h^3) \approx \frac{1}{2h}\{f(x+h) - f(x-h)\} + O(h^2)$$

Therefore, the 2nd order numerical approximation becomes

$$f'(x) \approx \frac{1}{2h}\{f(x+h) - f(x-h)\}$$

,which is called by the central difference formula

(Example) calculate $f'(1)$ for $f(x) = x^5$ with varying h : True value is $f'(1) = 5.0$

| h | $f'(1)$ with 1st order | $f'(1)$ with 2nd order |
|-------------|------------------------|------------------------|
| 1 | 31 | 16 |
| 0.5 | 13.1875 | 7.5625 |
| 0.25 | 8.20703125 | 5.62890625 |
| 0.125 | 6.416259766 | 5.156494141 |
| 0.0625 | 5.665298462 | 5.039077759 |
| 0.03125 | 5.322419167 | 5.009766579 |
| 0.015625 | 5.158710539 | 5.002441466 |
| 0.0078125 | 5.078737739 | 5.000610355 |
| 0.00390625 | 5.039215386 | 5.000152588 |
| 0.001953125 | 5.019569434 | 5.000038147 |

2. Definition of Jacobian and numerical approximation of Jacobian

Multi-variable function: $\mathbf{f}(\mathbf{x}) = 0$, $\mathbf{f} \in R^n$, $\mathbf{x} \in R^m$

$$\mathbf{f}(\mathbf{x}) = \begin{pmatrix} f_1(\mathbf{x}) \\ f_2(\mathbf{x}) \\ \vdots \\ f_n(\mathbf{x}) \end{pmatrix}, \quad \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{pmatrix}$$

$$\mathbf{f} \in R^2, \quad \mathbf{x} \in R^3$$

Example: $\mathbf{f}(x, y, z) = \begin{pmatrix} f_1(\mathbf{x}) \\ f_2(\mathbf{x}) \end{pmatrix} = \begin{pmatrix} x^2 + y^2 + z^2 + 2xy - 3yz \\ 3x - 2y + 5z \end{pmatrix}$

Definition of the Jacobian of the multi-variable function:

$$\frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}} = \begin{pmatrix} \frac{\partial f_1(\mathbf{x})}{\partial x_1} & \frac{\partial f_1(\mathbf{x})}{\partial x_2} & \dots & \frac{\partial f_1(\mathbf{x})}{\partial x_m} \\ \frac{\partial f_2(\mathbf{x})}{\partial x_1} & \frac{\partial f_2(\mathbf{x})}{\partial x_2} & \dots & \frac{\partial f_2(\mathbf{x})}{\partial x_m} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_n(\mathbf{x})}{\partial x_1} & \frac{\partial f_n(\mathbf{x})}{\partial x_2} & \dots & \frac{\partial f_n(\mathbf{x})}{\partial x_m} \end{pmatrix}$$

Example: $\frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}} = \begin{pmatrix} \frac{\partial f_1(\mathbf{x})}{\partial x_1} & \frac{\partial f_1(\mathbf{x})}{\partial x_2} & \frac{\partial f_1(\mathbf{x})}{\partial x_3} \\ \frac{\partial f_2(\mathbf{x})}{\partial x_1} & \frac{\partial f_2(\mathbf{x})}{\partial x_2} & \frac{\partial f_2(\mathbf{x})}{\partial x_3} \end{pmatrix} = \begin{pmatrix} 2x + 2y & 2y + 2x - 3z & 2z - 3y \\ 3 & -2 & 5 \end{pmatrix}$

Numerical approximation of the Jacobian using the finite difference formula

$$\frac{\partial f_k(\mathbf{x})}{\partial x_j} \approx \frac{f_k(x_1, \dots, x_{j-1}, x_j + \Delta x_j, x_{j+1}, \dots, x_m) - f_k(x_1, \dots, x_{j-1}, x_j - \Delta x_j, x_{j+1}, \dots, x_m)}{2\Delta x_j}$$

Remark: The formula has the perturbed values only for the x_j with $x_j \pm \Delta x_j$

(Example) Jacobian computing using the central difference formula

$$\mathbf{f}(x, y, z) = \begin{pmatrix} f_1(\mathbf{x}) \\ f_2(\mathbf{x}) \end{pmatrix} = \begin{pmatrix} x^2 + y^2 + z^2 + 2xy - 3yz \\ 3x - 2y + 5z \end{pmatrix} \text{ at } x = y = z = 1 \text{ with } \Delta x = \Delta y = \Delta z = 0.1$$

(i) Perturbation in x

Positive perturbation with $x = 1.1, y = z = 1$

$$\begin{pmatrix} f_1 \\ f_2 \end{pmatrix}_p = \begin{pmatrix} x^2 + y^2 + z^2 + 2xy - 3yz \\ 3x - 2y + 5z \end{pmatrix} = \begin{pmatrix} 2.4100 \\ 6.3000 \end{pmatrix}$$

Negative perturbation with $x = 0.9, y = z = 1$

$$\begin{pmatrix} f_1 \\ f_2 \end{pmatrix}_n = \begin{pmatrix} x^2 + y^2 + z^2 + 2xy - 3yz \\ 3x - 2y + 5z \end{pmatrix} = \begin{pmatrix} 1.6100 \\ 5.7000 \end{pmatrix}$$

Jacobian component due to variable x

$$\frac{\partial}{\partial x} \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} \approx \frac{1}{2 \times 0.1} \left\{ \begin{pmatrix} f_1 \\ f_2 \end{pmatrix}_p - \begin{pmatrix} f_1 \\ f_2 \end{pmatrix}_n \right\} = \frac{1}{2 \times 0.1} \left\{ \begin{pmatrix} 2.4100 \\ 6.3000 \end{pmatrix} - \begin{pmatrix} 1.6100 \\ 5.7000 \end{pmatrix} \right\} = \frac{1}{2 \times 0.1} \begin{pmatrix} 0.8000 \\ 0.6000 \end{pmatrix} = \begin{pmatrix} 4.0000 \\ 3.0000 \end{pmatrix}$$

(ii) Perturbation in y with the fixed values of $x = z = 1$

Jacobian component due to variable y

$$\frac{\partial}{\partial y} \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} \approx \frac{1}{2 \times 0.1} \left\{ \begin{pmatrix} f_1 \\ f_2 \end{pmatrix}_p - \begin{pmatrix} f_1 \\ f_2 \end{pmatrix}_n \right\} = \frac{1}{2 \times 0.1} \left\{ \begin{pmatrix} 2.1100 \\ 5.8000 \end{pmatrix} - \begin{pmatrix} 1.9100 \\ 6.2000 \end{pmatrix} \right\} = \begin{pmatrix} 1.0000 \\ -2.0000 \end{pmatrix}$$

(iii) Perturbation in z with the fixed values of $x = y = 1$

Jacobian component due to variable y

$$\frac{\partial}{\partial y} \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} \approx \frac{1}{2 \times 0.1} \left\{ \begin{pmatrix} f_1 \end{pmatrix} - \begin{pmatrix} f_1 \end{pmatrix} \right\} = \frac{1}{2 \times 0.1} \left\{ \begin{pmatrix} 1.9100 \\ 6.5000 \end{pmatrix} - \begin{pmatrix} 2.1100 \\ 5.5000 \end{pmatrix} \right\} = \begin{pmatrix} -1.0000 \\ 5.0000 \end{pmatrix}$$

(iv) Approximated Jacobian

$$\frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}} = \begin{pmatrix} 4 & 1 & -1 \\ 3 & -2 & 5 \end{pmatrix}$$

(v) Exact Jacobian

$$\frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}} = \begin{pmatrix} 2x+2y & 2y+2x-3z & 2z-3y \\ 3 & -2 & 5 \end{pmatrix} = \begin{pmatrix} 4 & 1 & -1 \\ 3 & -2 & 5 \end{pmatrix}$$

(Example) Jacobian computing using the forward difference formula

$$\mathbf{f}(x, y, z) = \begin{pmatrix} f_1(\mathbf{x}) \\ f_2(\mathbf{x}) \end{pmatrix} = \begin{pmatrix} x^2 + y^2 + z^2 + 2xy - 3yz \\ 3x - 2y + 5z \end{pmatrix} \text{ at } x = y = z = 1 \text{ with } \Delta x = \Delta y = \Delta z = 0.1$$

$$\mathbf{f}(1,1,1) = \begin{pmatrix} f_1(\mathbf{x}) \\ f_2(\mathbf{x}) \end{pmatrix} = \begin{pmatrix} 2 \\ 6 \end{pmatrix}$$

(Example): $\mathbf{f}(x, y, z) = \begin{pmatrix} f_1(\mathbf{x}) \\ f_2(\mathbf{x}) \end{pmatrix} = \begin{pmatrix} x^2 + y^2 + z^2 + 2xy - 3yz \\ 3x - 2y + 5z \end{pmatrix}$ at $x = y = z = 1$ with $\Delta x = \Delta y = \Delta z = 0.1$

(i) Perturbation in x

Jacobian component due to variable x

$$\frac{\partial}{\partial x} \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} \approx \frac{1}{0.1} \left\{ \begin{pmatrix} f_1 \\ f_2 \end{pmatrix}_p - \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} \right\} = \frac{1}{0.1} \left\{ \begin{pmatrix} 2.4100 \\ 6.3000 \end{pmatrix} - \begin{pmatrix} 2 \\ 6 \end{pmatrix} \right\} = \begin{pmatrix} 5.0 \\ 3.0 \end{pmatrix}$$

(ii) Perturbation in y with the fixed values of $x = z = 1$

$$\frac{\partial}{\partial y} \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} \approx \frac{1}{0.1} \left\{ \begin{pmatrix} f_1 \\ f_2 \end{pmatrix}_p - \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} \right\} = \frac{1}{0.1} \left\{ \begin{pmatrix} 2.1100 \\ 5.8000 \end{pmatrix} - \begin{pmatrix} 2 \\ 6 \end{pmatrix} \right\} = \begin{pmatrix} 1.1 \\ -2.0 \end{pmatrix}$$

(iii) Perturbation in z with the fixed values of $x = y = 1$

Jacobian component due to variable y

$$\frac{\partial}{\partial y} \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} \approx \frac{1}{0.1} \left\{ \begin{pmatrix} f_1 \\ f_2 \end{pmatrix}_p - \begin{pmatrix} f_1 \\ f_2 \end{pmatrix} \right\} = \frac{1}{0.1} \left\{ \begin{pmatrix} 1.9100 \\ 6.5000 \end{pmatrix} - \begin{pmatrix} 2 \\ 6 \end{pmatrix} \right\} = \begin{pmatrix} -0.9 \\ 5.0 \end{pmatrix}$$

(iv) Approximated Jacobian

$$\frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}} = \begin{pmatrix} 5.0 & 1.1 & -0.9 \\ 3.0 & -2.0 & 5.0 \end{pmatrix}$$

(v) Exact Jacobian

$$\frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}} = \begin{pmatrix} 2x+2y & 2y+2x-3z & 2z-3y \\ 3 & -2 & 5 \end{pmatrix} = \begin{pmatrix} 4 & 1 & -1 \\ 3 & -2 & 5 \end{pmatrix}$$

Tips

- 1) Use a small value for the perturbations $\Delta x, \Delta y, \Delta z$
- 2) Use the 2nd order central difference formula to enhance accuracy
- 3) Use the 1st order difference formula to save computing time with small values for the perturbations

3. Taylor series expansion of the multi-variable functions

(3-1) Two-variable scalar function $f(x, y) \in R$

$$f(x+h_x, y+h_y) \approx f(x, y) + \frac{\partial f}{\partial x} h_x + \frac{\partial f}{\partial y} h_y + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} h_x^2 + \frac{\partial^2 f}{\partial x \partial y} h_x h_y + \frac{1}{2} \frac{\partial^2 f}{\partial y^2} h_y^2 + O(h^3)$$

(3-2) Two-variable vector function $\mathbf{f}(x, y) = \begin{pmatrix} f_1(x, y) \\ f_2(x, y) \end{pmatrix} \in R^2$

$$\begin{aligned} \mathbf{f}(x+h_x, y+h_y) &= \begin{pmatrix} f_1(x+h_x, y+h_y) \\ f_2(x+h_x, y+h_y) \end{pmatrix} \\ &\approx \begin{pmatrix} f_1(x, y) + \frac{\partial f_1}{\partial x} h_x + \frac{\partial f_1}{\partial y} h_y \\ f_2(x, y) + \frac{\partial f_2}{\partial x} h_x + \frac{\partial f_2}{\partial y} h_y \end{pmatrix} = \begin{pmatrix} f_1(x, y) \\ f_2(x, y) \end{pmatrix} + \begin{pmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} \end{pmatrix} \begin{pmatrix} h_x \\ h_y \end{pmatrix} \\ &= \mathbf{f}(x, y) + \frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}} \Delta \mathbf{x} \leftarrow \frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}} = \begin{pmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} \end{pmatrix}, \quad \Delta \mathbf{x} = \begin{pmatrix} h_x \\ h_y \end{pmatrix} \end{aligned}$$

(3-3) Three-variable scalar function $f(x, y, z) \in R$

$$f(x+h_x, y+h_y, z+h_z) \approx f(x, y) + \frac{\partial f}{\partial x} h_x + \frac{\partial f}{\partial y} h_y + \frac{\partial f}{\partial z} h_z + O(h^2)$$

(3-4) Three-variable vector function $\mathbf{f}(x, y) = \begin{pmatrix} f_1(x, y) \\ f_2(x, y) \end{pmatrix} \in R^2$

$$\begin{aligned} \mathbf{f}(x+h_x, y+h_y, z+h_z) &= \begin{pmatrix} f_1(x+h_x, y+h_y, z+h_z) \\ f_2(x+h_x, y+h_y, z+h_z) \end{pmatrix} \\ &\approx \begin{pmatrix} f_1(x, y, z) + \frac{\partial f_1}{\partial x} h_x + \frac{\partial f_1}{\partial y} h_y + \frac{\partial f_1}{\partial z} h_z \\ f_2(x, y, z) + \frac{\partial f_2}{\partial x} h_x + \frac{\partial f_2}{\partial y} h_y + \frac{\partial f_2}{\partial z} h_z \end{pmatrix} \\ &= \begin{pmatrix} f_1(x, y, z) \\ f_2(x, y, z) \end{pmatrix} + \begin{pmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} & \frac{\partial f_1}{\partial z} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} & \frac{\partial f_2}{\partial z} \end{pmatrix} \begin{pmatrix} h_x \\ h_y \\ h_z \end{pmatrix} \\ &= \mathbf{f}(x, y, z) + \frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}} \Delta \mathbf{x} \leftarrow \frac{\partial \mathbf{f}(\mathbf{x})}{\partial \mathbf{x}} = \begin{pmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} & \frac{\partial f_1}{\partial z} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} & \frac{\partial f_2}{\partial z} \end{pmatrix}, \quad \Delta \mathbf{x} = \begin{pmatrix} h_x \\ h_y \\ h_z \end{pmatrix} \end{aligned}$$

4. Newton-Raphson Method: One of the most popular iterative method

- The 1st order Taylor series approximation of a function can be written as

$$f(x_{j+1}) \approx f(x_j) + f'(x)(x_{j+1} - x_j)$$

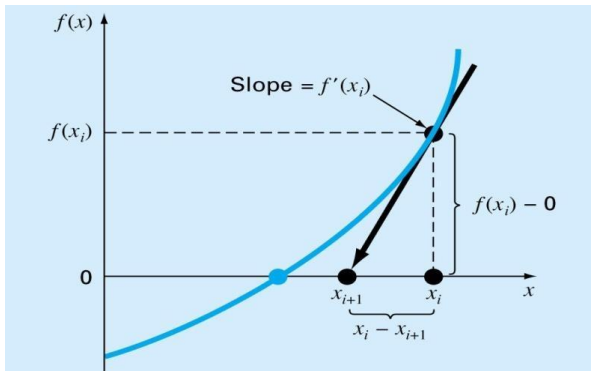
The Newton-Raphson method approximate the root with x_{j+1} satisfying $f(x_{j+1}) \approx f(x_j) + f'(x)(x_{j+1} - x_j) = 0$

Therefore,

$$\rightarrow f(x_j) + f'(x)(x_{j+1} - x_j) = 0$$

$$\rightarrow x_{j+1} - x_j = -\frac{f(x_j)}{f'(x_j)} \rightarrow x_{j+1} = x_j - \frac{f(x_j)}{f'(x_j)}$$

- Graphical depiction of the Newton-Raphson method



Function NEWTON(x, ITmax, h, epsilon)

!Pseudo code for Newton_Raphson method to find x satisfying $f(x)=0$ with a given initial point x

!Given variables: x(initial-value), h(small increment to calculate derivative)

! ITmax (number of maximum iteration allowed)

! Given tolerance: $\epsilon \ll 1$

! Given external function: $f(x)$

! Using numerical calculation of derivative information of the function $f(x)$

Do $j=1,2,3, \dots, ITmax$

! estimation of derivative using the central difference formula

```
!  
    fzero = f(x) xpl           !function value at zero perturbation  
    us = x+h fplus             ! positive perturbation  
    = f(xplus)                 ! function value at xplus  
  
!  
    xminus = x-h fmin          ! negative perturbation  
    us = f(xminus)             ! function value at xminus  
  
!  
    gradf = 0.5*(fplus-fminus)/h ! derivative(gradient) estimation
```

!Newton-Raphson method

x0=x

x \leftarrow x - fzero/gradf

! for the nonlinear system: gradf is a matrix

!convergence test

If abs(fzero) < epsilon, exit

!converged solution

If abs(x-x0)<epsilon, exit

!converged solution

!

End do NEW

TON = x

!similar to return value in C/C++

!

End NEWTON

5. The Secant Method

- In the Secant method the derivative is approximated using the 1st order finite difference formula with the following increment condition

$$x_j = x_{j-1} + h \rightarrow h = x_j - x_{j-1}$$

$$f'(x) \approx \frac{1}{h} \{f(x) - f(x-h)\} \rightarrow f'(x_j) \approx \frac{f(x_j) - f(x_{j-1})}{x_j - x_{j-1}}$$

Then the Newton-Raphson formula can be represented by the formula for the Secant Method

$$\begin{aligned} x_{j+1} &\approx x_j - \frac{f(x_j)}{f'(x_j)} \\ &= x_j - \frac{f(x_j)}{f(x_j) - f(x_{j-1})} (x_j - x_{j-1}) \end{aligned}$$

- Modified Secant method by directly using the backward difference formula (1st order)

$$f'(x) \approx \frac{1}{h} \{f(x) - f(x-h)\} \rightarrow f'(x_j) \approx \frac{f(x_j) - f(x_j - h)}{h}$$

$$\begin{aligned} x_{j+1} &\approx x_j - \frac{f(x_j)}{f'(x_j)} \\ &= x_j - \frac{hf(x_j)}{f(x_j) - f(x_j - h)} \end{aligned}$$

○ **Definition of open method**

- It needs functional information such as function value and its gradient at one points to find a root: Find \mathbf{x} satisfying the nonlinear equation $\mathbf{f}(\mathbf{x}) = 0$, $\mathbf{f} \in R^n$, $\mathbf{x} \in R^n$
- Bracketing methods are always convergent. However, the convergence of open methods highly depend on the initial estimation of the root, where function value and its gradient are calculated.

6. Newton-Raphson Method for the System of Nonlinear Equations

- Definition of the system of nonlinear equations

$\mathbf{f}(\mathbf{x}) = 0$, $\mathbf{f} \in R^n$, $\mathbf{x} \in R^n$, which has n unknowns $\mathbf{x} \in R^n$ and n nonlinear equations $\mathbf{f} \in R^n$

Expanded form

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{pmatrix} \in R^n, \quad \mathbf{f} = \begin{pmatrix} f_1(x_1, x_2, x_3, \dots, x_n) \\ f_2(x_1, x_2, x_3, \dots, x_n) \\ f_3(x_1, x_2, x_3, \dots, x_n) \\ \vdots \\ f_n(x_1, x_2, x_3, \dots, x_n) \end{pmatrix} = \mathbf{0} \in R^n$$

- Jacobean of the system of nonlinear equations

$$\frac{d\mathbf{f}}{d\mathbf{x}} = \mathbf{G} = \begin{pmatrix} \frac{df_1}{dx_1} & \frac{df_1}{dx_2} & \dots & \frac{df_1}{dx_n} \\ \frac{df_2}{dx_1} & \frac{df_2}{dx_2} & \dots & \frac{df_2}{dx_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{df_n}{dx_1} & \frac{df_n}{dx_2} & \dots & \frac{df_n}{dx_n} \end{pmatrix} \in R^{n \times n}$$

- 1st Order approximation of function value around \mathbf{x}

$$\begin{aligned}\mathbf{f}(\mathbf{x} + \mathbf{h}) &\approx \mathbf{f}(\mathbf{x}) + \frac{d\mathbf{f}}{d\mathbf{x}} \mathbf{h} \\ &\approx \mathbf{f}(\mathbf{x}) + \mathbf{G}\mathbf{h}\end{aligned}$$

- Newton-Raphson Method for the system of Nonlinear Equations

$$\mathbf{x}_{j+1} = \mathbf{x}_j + \mathbf{h}$$

The Newton-Raphson Method approximate the root with the vector satisfying

$$\begin{aligned}\mathbf{f}(\mathbf{x}_{j+1}) &\approx \mathbf{f}(\mathbf{x}_j) + \frac{d\mathbf{f}}{d\mathbf{x}}(\mathbf{x}_{j+1} - \mathbf{x}_j) \\ &\approx \mathbf{f}(\mathbf{x}_j) + \mathbf{G}(\mathbf{x}_{j+1} - \mathbf{x}_j)\end{aligned} \rightarrow \begin{aligned}\mathbf{f}(\mathbf{x}_j) + \mathbf{G}(\mathbf{x}_{j+1} - \mathbf{x}_j) &= 0 \\ \rightarrow \mathbf{G}\mathbf{x}_{j+1} &= \mathbf{G}\mathbf{x}_j - \mathbf{f}(\mathbf{x}_j) \\ \rightarrow \mathbf{x}_{j+1} &= \mathbf{G}^{-1}(\mathbf{G}\mathbf{x}_j - \mathbf{f}(\mathbf{x}_j)) \\ \rightarrow \mathbf{x}_{j+1} &= \mathbf{x}_j - \mathbf{G}^{-1}\mathbf{f}(\mathbf{x}_j)\end{aligned}$$

(Example) Newton-Raphson Method for the system of Nonlinear Equations

$$f_1(x, y) = x^2 + xy - 10 = 0$$

$$f_2(x, y) = 3xy^2 + y - 57 = 0$$

$$\frac{\partial f_1(x, y)}{\partial x} = 2x + y \quad \frac{\partial f_1(x, y)}{\partial y} = x$$

$$\frac{\partial f_2(x, y)}{\partial x} = 3y^2 \quad \frac{\partial f_2(x, y)}{\partial y} = 6xy + 1$$

$$\rightarrow \begin{matrix} \mathbf{G} = \begin{pmatrix} 2x + y & x \\ 3y^2 & 6xy + 1 \end{pmatrix} \\ \mathbf{G}^{-1} = \frac{1}{(2x + y)(6xy + 1) - 3xy^2} \begin{pmatrix} 6xy + 1 & -x \\ -3y^2 & 2x + y \end{pmatrix} \end{matrix} \quad \text{for } \mathbf{x} = \begin{pmatrix} x_j \\ y_j \end{pmatrix}$$

By using $\mathbf{x}_{j+1} = \mathbf{x}_j - \mathbf{G}^{-1}\mathbf{f}(\mathbf{x}_j)$

$$\begin{pmatrix} x_{j+1} \\ y_{j+1} \end{pmatrix} = \begin{pmatrix} x_j \\ y_j \end{pmatrix} - \frac{1}{(2x_j + y_j)(6x_j y_j + 1) - 3x_j y_j^2} \begin{pmatrix} 6x_j y_j + 1 & -x_j \\ -3y_j^2 & 2x_j + y_j \end{pmatrix} \begin{pmatrix} x_j^2 + x_j y_j - 10 \\ 3x_j y_j^2 + y_j - 57 \end{pmatrix}$$

Function NEWTON_SYS(n, x, ITmax, h, epsilon, error)

!-----

!Pseudo code for Newton_Raphson method to find x satisfying $f(x)=0$ with a given initial point x

!Given variables:

! n : (input) number of equations

! x(1:n): (input/output) initial-value

! ITmax:(input) maximum allowed iteration

! h(1:n): (input) small increments to calculate derivative

! epsilon: (input) given tolerance ($\ll 1$)

! error:(output) norm of function residual

!Be careful when we calculate the roots of nonlinear system of equations

! x(1:n), h(1:n), f(1:n), gradf(1:n, 1:n)

!-----

Do j=1,2,3,,ITmax

fzero(1:n) = f(x)

!function value at zero perturbation

! estimation of derivative using the central difference formula

do k = 1, n

xplus(1:n) = x(1:n)

! positive perturbation

xplus(k) = xplus(k) + h(k)

fplus(1:n) = f(xplus)

! function value at xplus

!

xminus(1:n) = x(1:n)

! negative perturbation

```

        xminus (k)    = xminus (k) - h(k)
        fminus (1:n)  = f(xminus)          ! function value at xminus
!
        gradf (1:n,k)= 0.5*(fplus(1:n)-fminus(1:n))/h(k) ! derivative(gradient) estimation
    end do
!Newton-Raphson method
        x0(1:n)=x(1:n)
        x(1:n) ← x(1:n) – (gradf)-1 *fzero(1:n)    ! for the nonlinear system: gradf is a matrix
!convergence test
        If norm(fzero) < epsilon, exit          !converged solution
        If norm(x-x0)<epsilon, exit              !converged solution
    End do
!
    error = norm(f(x)) ! residual in function value
!
End NEWTON_SYS

```

Appendix: Problem set for the System of Nonlinear Equations

General Problem Statements

$$\mathbf{f}(\mathbf{x}) = \mathbf{0}, \quad \mathbf{f}, \mathbf{x} \in R^n \rightarrow \mathbf{f} = \begin{pmatrix} f_1(x_1, x_2, \dots, x_n) \\ f_2(x_1, x_2, \dots, x_n) \\ \vdots \\ f_n(x_1, x_2, \dots, x_n) \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \quad \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

1. Problem #1

$$\mathbf{f} = \begin{pmatrix} x_1 + 3\ln(x_1) - x_2^2 \\ 2x_1^2 - x_1x_2 - 5x_1 + 1.0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 1.0 \\ -2.0 \end{pmatrix}$$

2. Problem #2

$$\mathbf{f} = \begin{pmatrix} x_1^2 + x_1x_2^2 - 9 \\ 3x_1^2x_2 - x_2^3 - 4 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 1.2 \\ 2.5 \end{pmatrix}$$

3. Problem #3

$$\mathbf{f} = \begin{pmatrix} x_1 + 2x_2 - 3.0 \\ 2x_1^2 + x_2^2 - 5.0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 1.5 \\ 1.0 \end{pmatrix}$$

4. Problem #4

$$\mathbf{f} = \begin{pmatrix} 3x_1^2 + 4x_2^2 - 1.0 \\ x_2^3 - 8x_1^3 - 1.0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} -0.5 \\ 0.25 \end{pmatrix}$$

5. Problem #5

$$\mathbf{f} = \begin{pmatrix} 4x_1^2 + x_2^2 - 4.0 \\ x_1 + x_2 - \sin(x_1 - x_2) \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 1.0 \\ 0.0 \end{pmatrix}$$

6. Problem #6 : Not converged

$$\mathbf{f} = \begin{pmatrix} x_1^5 + x_2^3 x_3^4 + 1.0 \\ x_1^2 x_2 x_3 \\ x_3^4 - 1.0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} -10 \\ -10 \\ -10 \end{pmatrix}$$

7. Problem #7

$$\mathbf{f} = \begin{pmatrix} x_1^2 + x_2 - 37 \\ x_1 - x_2^2 - 5.0 \\ x_1 + x_2 + x_3 - 3.0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 5.0 \\ 0.0 \\ -2.0 \end{pmatrix}$$

8. Problem #8

$$\mathbf{f} = \begin{pmatrix} 12x_1 - 3x_2^2 - 4x_3 - 7.17 \\ x_1^2 + 10x_2 - x_3 - 11.54 \\ x_2^3 + 7x_3 - 7.631 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 3.0 \\ 0.0 \\ 1.0 \end{pmatrix}$$

9. Problem #9

$$\mathbf{f} = \begin{pmatrix} 10x_2 - 10x_1^2 \\ 1 - x_1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} -1.2 \\ 1.0 \\ 0 \end{pmatrix}$$

10. Problem #10: Not converged

$$\mathbf{f} = \begin{pmatrix} -13.0 + x_1 - x_2^3 + 5x_2^2 - 2x_2 \\ -29.0 + x_1 + x_2^3 + x_2^2 - 14x_2 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 15.0 \\ 1.0 \\ 0 \end{pmatrix}$$

11. Problem #11

$$\mathbf{f} = \begin{pmatrix} 10.0x_2 - 10x_1^2 \\ 1.0 - x_1 \\ 10.0x_4 - 10x_3^2 \\ 1.0 - x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 0.5 \\ 0.5 \\ 0.5 \\ 0.5 \end{pmatrix}$$

12.Problem #12 : Poor convergence

$$\mathbf{f} = \begin{pmatrix} x_1 + 10x_2 \\ \sqrt{5}x_3 - \sqrt{5}x_4 \\ (x_2 - 2x_3)^2 \\ \sqrt{10}(x_1 - x_4) \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 0.5 \\ 0.5 \\ 0.5 \\ 0.5 \end{pmatrix}$$

13.Problem #13 : Poor convergence

$$\mathbf{f} = \begin{pmatrix} x_1^2 - x_2 - 1 \\ x_2^2 - x_1 - 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 0.5 \\ 0.5 \end{pmatrix}$$

14.Problem #14 : Poor convergence

$$\mathbf{f} = \begin{pmatrix} x_1 - x_2^2 \\ (x_2 - 1)^2(x_2 - 2)^2 + (x_1 - x_2^2)^2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \mathbf{x}_0 = \begin{pmatrix} 0.5 \\ 0.5 \end{pmatrix}$$

```

!-----
! Problem Set for Nonlinear Algebraic Equations
!-----
SUBROUTINE NAE_Problems(IND_PROBLEM,IND_case,No_variable,X,Fun,No_fun_call);
!-----
IMPLICIT DOUBLE PRECISION (A-H,O-Z)
!-----
! Input:
!     IND_PROBLEM: Problem Number
!     IND_Case
!         = 1: initialization routine
!         = 2: evaluation of function vectors
!-----
DIMENSION X(*),Fun(*)
!-----
IF(IND_case==2) No_fun_call = No_fun_call + 1
!-----
NSELECT = 0
SELECT CASE (IND_PROBLEM)
!-----
CASE (1);
!-----
    IF(IND_Case==1) THEN
        No_variable = 2;
!
        X(1) = 1.0
        X(2) = -2.0
        NSELECT = 1;
!
    ELSEIF(IND_Case==2) THEN
        Fun(1) = X(1) + 3.0*LOG(X(1))-X(2)*X(2)
        Fun(2) = 2*X(1)*X(1) - X(1)*X(2) - 5.0*X(1) + 1.0
        NSELECT = 1;
    END IF
!-----

```



```

!-----
CASE (2) ;
!-----
    IF(IND_Case==1) THEN
        No_variable = 2;
!
        X(1) = 1.2
        X(2) = 2.5
        NSELECT = 1;
!-----
    ELSEIF(IND_Case==2) THEN
        Fun(1) = X(1)*X(1) + X(1)*X(2)*X(2) - 9.0
        Fun(2) = 3*X(1)*X(1)*X(2) - X(2)**3 - 4.0
        NSELECT = 1;
    END IF
!-----

!-----
CASE (3) ;
!-----
    IF(IND_Case==1) THEN
        No_variable = 2;
!
        X(1) = 1.5
        X(2) = 1.0
        NSELECT = 1;
!-----
    ELSEIF(IND_Case==2) THEN
        Fun(1) = X(1) + 2.0*X(2) - 3.0
        Fun(2) = 2*X(1)*X(1) + X(2)*X(2) - 5.0

```

```
NSELECT = 1;  
END IF
```

```
!  
!  
CASE (4) ;
```

```
!  
    IF(IND_Case==1) THEN
```

```
        No_variable = 2;
```

```
        X(1) = -0.5
```

```
        X(2) = 0.25
```

```
        NSELECT = 1;
```

```
    ELSEIF(IND_Case==2) THEN
```

```
        Fun(1) = 3.0*X(1)*X(1) + 4*X(2)*X(2) - 1.0
```

```
        Fun(2) = X(2)**3 - 8.0*X(1)**3 - 1.0
```

```
        NSELECT = 1;
```

```
    END IF
```

```
!  
!  
CASE (5) ;
```

```
!  
    IF(IND_Case==1) THEN
```

```
        No_variable = 2;
```

```
        X(1) = 1.0
```

```
        X(2) = 0.0
```

```
        NSELECT = 1;
```

```

ELSEIF(IND_Case==2) THEN
    Fun(1) = 4.0*X(1)*X(1) + X(2)*X(2) - 4.0
    Fun(2) = X(1) + X(2) - SIN(X(1)-X(2))
    NSELECT = 1;
END IF

```

!-----

!-----

```

CASE (6) ;

```

!-----

```

IF(IND_Case==1) THEN
    No_variable = 3;

```

!

```

    X(1) = -10.0
    X(2) = -10.0
    X(3) = -10.0
    NSELECT = 1;

```

!-----

```

ELSEIF(IND_Case==2) THEN
    Fun(1) = X(1)**5 + (X(2)**3)*(X(3)**4) + 1.0
    Fun(2) = X(1)*X(1)*X(2)*X(3)
    Fun(3) = X(3)**4 - 1.0 NSELE
    CT = 1;

```

```

END IF

```

!-----

!-----

```

CASE (7) ;

```

!-----

```

IF(IND_Case==1) THEN
    No_variable = 3;

```

```

X(1)= 5.0
X(2)= 0.0
X(3)= -2.0
NSELECT = 1;

```

```

ELSEIF(IND_Case==2) THEN
    Fun(1)= X(1)**2 + X(2) - 37.0
    Fun(2)= X(1) - X(2)*X(2) - 5.0
    Fun(3)= X(1) + X(2) + X(3) - 3.0
    NSELECT = 1;
END IF

```

```

CASE (8);

```

```

IF(IND_Case==1) THEN
    No_variable = 3;

```

```

X(1)= 3.0
X(2)= 0.0
X(3)= 1.0
NSELECT = 1;

```

```

ELSEIF(IND_Case==2) THEN
    Fun(1)= 12.0*X(1) - 3.0*X(2)**2 - 4.0*X(3) - 7.17
    Fun(2)= X(1)**2 + 10.0*X(2) - X(3) - 11.54
    Fun(3)= X(2)**3 + 7.0*X(3) - 7.631
    NSELECT = 1;
END IF

```

```
!-----
CASE (9) ; ! Ref Broyden A Class of Methods for Solving Nonlinear Simultaneous Equations
!-----
```

```
  IF(IND_Case==1) THEN
    No_variable = 2;
!
    X(1) = -1.2
    X(2) =  1.0
!
    NSELECT = 1;
!-----
  ELSEIF(IND_Case==2) THEN Fu
    n(1) = 10.0*(X(2) - X(1)**2) F
    un(2) = 1.0 - X(1)
!
    NSELECT = 1;
  END IF
!-----
```

```
!-----
CASE (10) ; ! Ref Broyden A Class of Methods for Solving Nonlinear Simultaneous Equations
!-----
```

```
  IF(IND_Case==1) THEN
    No_variable = 2;
!
    X(1) =  15.0
    X(2) =  1.0
!
    NSELECT = 1;
!-----
  ELSEIF(IND_Case==2) THEN
    Fun(1) = -13.0 + X(1) + ((-X(2) + 5.0)*X(2) - 2.0)*X(2)
    Fun(2) = -29.0 + X(1) + (( X(2) + 1.0)*X(2) - 14.0)*X(2)
```

```

!
      NSELECT = 1;
ENDIF
!-----
!-----
CASE (11) ; ! Ref Shanghai Multi-step Nonlinear ABS Methods and Their Efficiency Analysis 1991 (Extended Rosenbloack function)
!-----
      IF(IND_Case==1) THEN
        No_variable = 4;
!
        X(1:4) = 0.5
!
        NSELECT = 1;
!-----
      ELSEIF(IND_Case==2) THEN Fu
        n(1) = 10.0*(X(2) - X(1)**2) F
        un(2) = 1.0 - X(1)
        Fun(3) = 10.0*(X(4) - X(3)**2)
        Fun(4) = 1.0 - X(3)
!
        NSELECT = 1;
      ENDIF
!-----
!-----
CASE (12) ; !! Ref Shanghai Multi-step Nonlinear ABS Methods and Their Efficiency Analysis 1991 (Extended Powell singular function)
!-----
      IF(IND_Case==1) THEN
        No_variable = 4;
!
        X(1:4) = 0.5
!
        NSELECT = 1;
!-----

```

```

ELSEIF(IND_Case==2) THEN
  Fun(1) = X(1) + 10.0*X(2)
  Fun(2) = Sqrt(5.0)*(X(3) - X(4))
  Fun(3) = (X(2) - 2.0*X(3))**2
  Fun(4) = Sqrt(10.0)*(X(1) - X(4))**2

```

```

!
  NSELECT = 1;

```

```

END IF

```

```

!-----
CASE (13); !! Ref Atluri, A Modified Newton Method for Solving NAEs Example #1
!-----

```

```

IF(IND_Case==1) THEN

```

```

  No_variable = 2;

```

```

  X(1:2) = 0.5

```

```

  NSELECT = 1;

```

```

ELSEIF(IND_Case==2) THEN

```

```

  Fun(1) = X(1)**2 - X(2)-1.0

```

```

  Fun(2) = X(2)**2 - X(1)-1.0

```

```

  NSELECT = 1;

```

```

END IF

```

```

!-----
CASE (14); !! Ref Atluri, A Modified Newton Method for Solving NAEs Example #2
!-----

```

```

IF(IND_Case==1) THEN

```

```

  No_variable = 2;

```

```

!
      X(1:2)=0.5
!
      NSELECT = 1;
!-----
      ELSEIF(IND_Case==2) THEN
        Fun(1)= X(1) - X(2)**2
        Fun(2)= ((X(2)-1.0)**2)*((X(2)-2.0)**2) + (X(1)-X(2)**2)**2
!
        NSELECT = 1;
      END IF
!-----

!-----
CASE (15) ; ! ! Ref Atluri, A Modified Newton Method for Solving NAEs Example #3
!-----
      a1 = 25.0;   b1 = 1.0;   c1 = 2.0;
      a2 = 3.0;   b2 = 4.0;   c2 = 5.0;
      IF(IND_Case==1) THEN
        No_variable = 2;
!
        X(1)=-10.0
        X(2)= -1.0
!
        NSELECT = 1;
!-----
      ELSEIF(IND_Case==2) THEN
        Fun(1)= X(1)**3 -3.0*X(1)*X(2)**2 +a1*(2.0*X(1)**2 +      X(1)*X(2)) + b1*X(2)**2 + c1*X(1) + a2*X(2)
        Fun(2)= -X(2)**3 +3.0*X(2)*X(1)**2 +a1*(      X(2)**2 -4.0*X(1)*X(2)) + b2*X(1)**2 + c2
!
        NSELECT = 1;
      END IF
!-----

```



```

!-----
CASE (16) ; ! ! Ref Atluri, A Modified Newton Method for Solving NAEs Example #4
!-----
IF(IND_Case==1) THEN
    No_variable = 3;
!
    X(1:3) = 0.1
!
    NSELECT = 1;
!-----
ELSEIF(IND_Case==2) THEN
    Fun(1) = X(1) + X(2) + X(3) - 3.0
    Fun(2) = X(1)*X(2) + 2.0*X(2)**2 + 4.0*X(3)**2 - 7.0
    Fun(3) = X(1)**8 + X(2)**4 + X(3)**9 - 3.0
!
    NSELECT = 1;
END IF
!-----

```

```

!-----
!-----
! Large Scale Problem by Adjusting N the number of equations
!-----
!-----
CASE (101) ; ! Ref An Autoadattative limited memory Broyden's Method to Solve Systems of NEs :: Broyden banded function
!-----
    N = 10
    IF(IND_Case==1) THEN
        No_variable = N;
!

```

```

X(1:N)= 0.0
NSELECT= 1;
!-----
ELSEIF(IND_Case==2) THEN
  Fun(1)= X(1)*(2.0+5.0*X(1)**2)+1.0 - X(2)*(1.0+X(2))
  Fun(2)= X(2)*(2.0+5.0*X(2)**2)+1.0 - X(1)*(1.0+X(1)) - X(3)*(1.0+X(3))
  Fun(3)= X(3)*(2.0+5.0*X(3)**2)+1.0 - X(1)*(1.0+X(1)) - X(2)*(1.0+X(2)) - X(4)*(1.0+X(4))
  Fun(4)= X(4)*(2.0+5.0*X(4)**2)+1.0 - X(1)*(1.0+X(1)) - X(2)*(1.0+X(2)) - X(3)*(1.0+X(3)) - X(5)*(1.0+X(5))
  Fun(5)= X(5)*(2.0+5.0*X(5)**2)+1.0 - X(1)*(1.0+X(1)) - X(2)*(1.0+X(2)) - X(3)*(1.0+X(3)) - X(4)*(1.0+X(4)) - X(6)*(1.0+X(6))
!
  DO J = 6, N-1
    Fun(J)= X(J)*(2.0+5.0*X(J)**2)+1.0
    DO K = J-5, J-1
      Fun(J)= Fun(J) - X(K)*(1.0+X(K))
    END DO
    Fun(J)= Fun(J) - X(J+1)*(1.0+X(J+1))
  END DO
!
  Fun(N)= X(N)*(2.0+5.0*X(N)**2)+1.0
  DO K = N-5, N-1
    Fun(N)= Fun(N) - X(K)*(1.0+X(K))
  END DO
!
  NSELECT= 1;
END IF
!-----

!-----
CASE (102) ; ! Ref An Autoadattive limited memory Broyden's Method to Solve Systems of NEs :: Martinez function
!-----
  N= 10
  IF(IND_Case==1) THEN
    No_variable= N;
  !

```

```

X(1:N)= 0.0
NSELECT = 1;
!-----
ELSEIF(IND_Case==2) THEN
  Fun(1) = (3.0 - 0.1*X(1))*X(1) + 1.0 - 2.0*X(2) + X(1)
  DO J = 2, N-1
    Fun(J) = (3.0 - 0.1*X(J))*X(J) + 1.0 - X(J-1) - 2.0*X(J+1) + X(J)
  ENDDO
  Fun(N) = (3.0 - 0.1*X(N))*X(N) + 1.0 - 2.0*X(N-1) + X(N)
!
  NSELECT = 1;
END IF
!-----

!-----
CASE (103); ! Ref An Autoadattative limited memory Broyden's Method to Solve Systems of NEs :: Broyden tridiagonal function
!-----
N = 10
IF(IND_Case==1) THEN
  No_variable = N;
!
  X(1:N)= 0.0
  NSELECT = 1;
!-----
ELSEIF(IND_Case==2) THEN
  Fun(1) = (3.0 - 2.0*X(1))*X(1) + 1.0 - 2.0*X(2)
  Fun(N) = (3.0 - 2.0*X(N))*X(N) + 1.0 - X(N-1)
  DO J = 2, N-1
    Fun(J) = (3.0 - 2.0*X(J))*X(J) + 1.0 - X(J-1) - 2.0*X(J+1)
  ENDDO
!
  NSELECT = 1;
END IF
!-----

```

```

!-----
CASE (104); ! Ref An Autoadatative limited memory Broyden's Method to Solve Systems of NEs :: Spedicato function 4
!-----
      N = 10
      IF(IND_Case==1) THEN
        No_variable = N;

!
        X(1:N-1) = -1.2
        X(N)      = 1.0
        NSELECT   = 1;

!-----
      ELSEIF(IND_Case==2) THEN
        K = 0
        DO J = 1, N
          K = K + 1
          IF(K.EQ.1) THEN ; ! Odd case of J F
            un(J) = 1.0 - X(J)
          ELSE ; ! Even case of J
            Fun(J) = 100.0*(X(J) - X(J-1)**2)
            K = 0
          END IF
        END DO

!
        NSELECT = 1;
      END IF

!-----

!-----
CASE (105); ! Ref An Autoadaptive limited memory Broyden's Method to Solve Systems of NEs :: Discrete integral equation function
!-----
      N = 10
      H = 1.0/FLOAT(N+1)

```

```

HH= 0.5*H
IF(IND_Case==1) THEN
    No_variable = N;
!
    DO J = 1, N
        TJ = H*FLOAT(J)
        X(J) = TJ*(TJ-1.0)
    END DO
!
    NSELECT = 1;
!-----
ELSEIF(IND_Case==2) THEN
    DO J = 1, N
        TJ = H*FLOAT(J)
!
        Sum1 = 0.0
        DO K = 1, J
            TK = H*FLOAT(K)
            Sum1 = Sum1 + TK*(X(K)+TK+1.0)**3
        END DO
!
        Sum2 = 0.0
        DO K = J+1, N
            TK = H*FLOAT(K)
            Sum2 = Sum2 + (1.0 - TK)*(X(J)+TK+1.0)**3
        END DO
!
        Fun(J) = X(J) + HH*((1.0- TJ)*Sum1 + TJ*sum2)
    END DO
!
    NSELECT = 1;
END IF
!-----

```

```
!-----
CASE (106) ; ! Ref Shanghai Multi-step Nonlinear ABS Methods and Their Efficiency Analysis 1991 (Brown Problem)
!-----
```

```
    N = 10
    IF(IND_Case==1) THEN
        No_variable = N;
```

```
    X(1:N) = 0.5
```

```
    NSELECT = 1;
```

```
!-----
ELSEIF(IND_Case==2) THEN
```

```
    Pr1 = 1.0
```

```
    DO K = 1, N
```

```
        Pr1 = Pr1*X(K)
```

```
    END DO
```

```
    Fun(N) = -1.0 + Pr1
```

```
    DO J = 2, N-1
```

```
        Fun(J) = -FLOAT(N+1)
```

```
        DO K = J+1, N
```

```
            Fun(J) = Fun(J) + X(K)
```

```
        END DO
```

```
    END DO
```

```
    NSELECT = 1;
```

```
    END IF
```

```
!-----
CASE (107) ; ! ! Ref Aituri, A Modified Newton Method for Solving NAEs Example #5
!-----
```

```
    N = 10
```

```
    X0 = 0.0
```

```

XN= 20.0
IF(IND_Case==1) THEN
    No_variable = N;
    X(1:N)= 1.0
!
    NSELECT = 1;
!-----
    ELSEIF(IND_Case==2) THEN
        DO J = 1, N
            IF(J==1) THEN
                Xm = X0
                Xx = X(J)
                Xp = X(J+1)
            ELSE IF(J==N) THEN
                Xm = X(J-1)
                Xx = X(J) X
                p = XN
            ELSE
                Xm = X(J-1)
                Xx = X(J) X
                p = X(J+1)
            END IF
!
            Fun(J) = 3.0*Xx*(Xp -2.0*Xx + Xm) + 0.25*(Xp-Xm)**2
        END DO
!
        NSELECT = 1;
    END IF
!-----
!-----
CASE (108) ; !! Ref Atluri, A Modified Newton Method for Solving NAEs Example #6
!-----
    IF(IND_Case==1) THEN

```

```

        No_variable = 10;
        X(1:N) = -0.1
!
        NSELECT = 1;
!-----
        ELSEIF(IND_Case==2) THEN
            Fun(1) = (3.0-5.0*X(1))*X(1) + 1.0 -2.0*X(2)
            Fun(10) = (3.0-5.0*X(10))*X(10) + 1.0 - X(9)
            DO J = 2, 9
                Fun(J) = (3.0-5.0*X(J))*X(J) - X(J-1) - 2.0*X(J+1)
            END DO
!
            NSELECT = 1;
        END IF
!-----

!-----
CASE (109) ; !! Ref Atluri, A Modified Newton Method for Solving NAEs Example #7
!-----
        N = 10
        H = 1.0/FLOAT(N+1)
        H2 = 1.0/H**2
!
        X0 = 4.0
        XN = 1.0
!
        IF(IND_Case==1) THEN
            No_variable = N;
            X(1:N) = 0.5
!
            NSELECT = 1;
!-----
        ELSEIF(IND_Case==2) THEN
            DO J = 1, N

```



```

      IF(J==1) THEN
        Xm = X0
        Xx = X(J)
        Xp = X(J+1)
      ELSE IF(J==N) THEN
        Xm = X(J-1)
        Xx = X(J)
        p = XN
      ELSE
        Xm = X(J-1)
        Xx = X(J)
        p = X(J+1)
      END IF
!
      Fun(J) = H2*(Xp -2.0*Xx + Xm) - 1.5*Xx**2
    END DO
!
      NSELECT = 1;
    END IF
!-----

!-----
  END SELECT
!
  IF(NSELECT.EQ.0) THEN
    PRINT*, 'NO PROBLEM IS SELECTED, SEE SUBROUTINE NLP_Problems for IND_PROBLEM = ', IND_PROBLEM
    STOP
  END IF
!
RETURN
END
!=====

```

| Np | lt_newt | Nf_newt | Fn_newt | Dx_newt | lt_brdn | Nf_brdn | Fn_brdn | Dx_Brdn | lt_Tmas | Nf_Tmas | Fn_Tmas | Dx_Tmas | lt_mart | Nf_mart | Fn_mart | Dx_mart | | | | |
|----|---------|---------|--------------|--------------|---------|---------|--------------|--------------|---------|---------|--------------|--------------|---------|---------|--------------|--------------|-----|-----|--------------|--------------|
| 1 | 11 | 55 | 0.314018E-15 | 0.728109E-16 | 14 | 14 | 0.210486E-13 | 0.490187E-14 | 19 | 19 | 0.289978E-12 | 0.135531E-12 | 15 | 15 | 0.378128E-14 | 0.127834E-14 | 500 | 505 | 0.173576E-02 | 0.616325E-05 |
| 2 | 10 | 50 | 0.237783E-12 | 0.217161E-13 | 14 | 14 | 0.120334E-12 | 0.118316E-13 | 13 | 13 | 0.564844E-12 | 0.596211E-13 | 14 | 14 | 0.118498E-13 | 0.103636E-14 | 500 | 505 | 0.587316E-02 | 0.555262E-05 |
| 3 | 10 | 50 | 0.439626E-14 | 0.945924E-15 | 13 | 13 | 0.595989E-12 | 0.221141E-12 | 12 | 12 | 0.106766E-13 | 0.229617E-14 | 15 | 15 | 0.309272E-14 | 0.154266E-14 | 500 | 505 | 0.397494E-03 | 0.219393E-05 |
| 4 | 10 | 50 | 0.157009E-15 | 0.493563E-16 | 20 | 20 | 0.388806E-12 | 0.734353E-13 | 11 | 11 | 0.551088E-12 | 0.205790E-12 | 11 | 11 | 0.414695E-13 | 0.662863E-14 | 500 | 505 | 0.577830E-05 | 0.498691E-07 |
| 5 | 10 | 50 | 0.157009E-15 | 0.107252E-15 | 12 | 12 | 0.415029E-12 | 0.283751E-12 | 13 | 13 | 0.180706E-13 | 0.301705E-14 | 13 | 13 | 0.351984E-13 | 0.444482E-14 | 500 | 505 | 0.943471E-04 | 0.981775E-06 |
| 6 | 35 | 245 | 0.963470E-12 | 0.372914E-06 | 101 | 101 | 0.230735E+85 | 0.788668E+10 | 101 | 101 | NaN | NaN 505 | 101 | 101 | 0.194128E+27 | 0.102558E-13 | 500 | NaN | NaN | |
| 7 | 11 | 77 | 0.000000E+00 | 0.000000E+00 | 16 | 16 | 0.876837E-12 | 0.235614E-12 | 17 | 17 | 0.637615E-13 | 0.193977E-13 | 14 | 14 | 0.705456E-12 | 0.319848E-12 | 500 | 505 | 0.526664E-02 | 0.873318E-05 |
| 8 | 11 | 77 | 0.725195E-15 | 0.702075E-16 | 17 | 17 | 0.211677E-13 | 0.197035E-14 | 15 | 15 | 0.957971E-14 | 0.109693E-14 | 18 | 18 | 0.644159E-13 | 0.736436E-14 | 500 | 505 | 0.110273E-01 | 0.127884E-04 |
| 9 | 10 | 50 | 0.000000E+00 | 0.000000E+00 | 16 | 16 | 0.392523E-14 | 0.395443E-15 | 13 | 13 | 0.785046E-15 | 0.784943E-16 | 19 | 19 | 0.628086E-14 | 0.467219E-15 | 500 | 505 | 0.155750E-02 | 0.459616E-04 |
| 10 | 101 | 500 | 0.784772E+01 | 0.121673E+02 | 101 | 101 | 0.177408E+02 | 0.136895E+02 | 101 | 101 | 0.924939E+01 | 0.148465E+02 | 101 | 101 | 0.385372E+03 | 0.615735E+02 | 500 | 505 | NaN | NaN |
| 11 | 9 | 81 | 0.000000E+00 | 0.000000E+00 | 77 | 77 | 0.209302E-12 | 0.162038E-13 | 20 | 20 | 0.858842E-12 | 0.864412E-13 | 41 | 41 | 0.111022E-14 | 0.117153E-15 | 500 | 505 | 0.159514E-02 | 0.942575E-05 |
| 12 | 101 | 900 | NaN | NaN | 101 | 101 | 0.611603E-04 | 0.280025E-02 | 101 | 101 | 0.667863E-06 | 0.232192E-03 | 101 | 101 | 0.125840E-04 | 0.130658E-02 | 500 | 505 | 0.391605E-02 | 0.201710E-04 |
| 13 | 101 | 500 | NaN | NaN | 12 | 12 | 0.239142E-12 | 0.106927E-12 | 12 | 12 | 0.239142E-12 | 0.106927E-12 | 12 | 12 | 0.155431E-14 | 0.695109E-15 | 500 | 505 | 0.426203E-02 | 0.289752E-04 |
| 14 | 101 | 500 | 0.826539E-11 | 0.366395E-05 | 78 | 78 | 0.982218E-12 | 0.300409E-07 | 101 | 101 | 0.997873E-10 | 0.211875E-06 | 35 | 35 | 0.405667E-12 | 0.386207E-06 | 500 | 505 | 0.416466E-03 | 0.258069E-03 |
| 15 | 101 | 500 | 0.447148E+01 | 0.294290E+00 | 101 | 101 | 0.409754E+01 | 0.150103E-02 | 101 | 101 | 0.198576E+02 | 0.191478E+00 | 101 | 101 | 0.334405E+04 | 0.548425E+01 | 500 | 505 | 0.116989E+09 | 0.193125E+01 |
| 16 | 30 | 210 | 0.904843E-12 | 0.271993E-12 | 89 | 89 | 0.173068E-12 | 0.612027E-13 | 23 | 23 | 0.213220E-12 | 0.675918E-13 | 41 | 41 | 0.317583E-12 | 0.677076E-13 | 500 | 505 | 0.419002E-02 | 0.251980E-04 |
| 10 | 17 | 357 | 0.276556E-12 | 0.376083E-13 | 101 | 101 | 0.919831E+13 | 0.132828E+04 | 101 | 101 | 0.165520E+00 | 0.317172E-01 | 101 | 101 | 0.835198E+01 | 0.296900E+00 | 500 | 505 | 0.728118E-01 | 0.130043E-03 |
| 1 | 10 | 210 | 0.289468E-14 | 0.212347E-14 | 34 | 34 | 0.900751E-12 | 0.346585E-12 | 22 | 22 | 0.340257E-13 | 0.133692E-13 | 30 | 30 | 0.314446E-12 | 0.866142E-13 | 500 | 505 | NaN | NaN |
| 2 | 11 | 231 | 0.352834E-15 | 0.764035E-16 | 48 | 48 | 0.853439E-12 | 0.297119E-12 | 23 | 23 | 0.159601E-12 | 0.396714E-13 | 34 | 34 | 0.271327E-12 | 0.495800E-13 | 500 | 505 | NaN | NaN |
| 3 | 9 | 189 | 0.000000E+00 | 0.000000E+00 | 101 | 101 | 0.192036E-03 | 0.289833E-05 | 39 | 39 | 0.306079E-13 | 0.380775E-15 | 30 | 30 | 0.213685E-12 | 0.414231E-14 | 500 | 505 | 0.594295E-01 | 0.480422E-04 |
| 4 | 10 | 210 | 0.202302E-16 | 0.173801E-16 | 19 | 19 | 0.482188E-12 | 0.479602E-12 | 18 | 18 | 0.315927E-12 | 0.312288E-12 | 16 | 16 | 0.303084E-12 | 0.307928E-12 | 500 | 505 | 0.700192E-04 | 0.813046E-06 |
| 5 | 101 | 2100 | NaN | NaN | 101 | 101 | 0.316228E+00 | 0.123955E-13 | 101 | 101 | 0.316228E+00 | 0.471344E-07 | 101 | 101 | 0.316228E+00 | 0.931649E-12 | 500 | 505 | 0.316749E+00 | 0.361314E-02 |
| 6 | 22 | 462 | 0.315274E-12 | 0.426300E-10 | 101 | 101 | 0.553139E+12 | 0.145194E+06 | 101 | 101 | 0.136526E-03 | 0.260155E-01 | 101 | 101 | 0.207699E-01 | 0.181703E-01 | 500 | 505 | 0.114560E-01 | 0.146905E-03 |
| 7 | 11 | 231 | 0.464440E-16 | 0.938605E-17 | 33 | 33 | 0.724831E-12 | 0.216322E-12 | 22 | 22 | 0.650093E-12 | 0.150471E-12 | 31 | 31 | 0.169978E-12 | 0.382370E-13 | 500 | 505 | 0.298611E-03 | 0.918865E-06 |
| 8 | 11 | 231 | 0.333990E-13 | 0.202870E-15 | 101 | 101 | 0.245979E+04 | 0.117144E+02 | 34 | 34 | 0.330289E-12 | 0.113646E-13 | 101 | 101 | 0.637111E+05 | 0.157089E+03 | 500 | 505 | 0.171403E+01 | 0.871518E-04 |