#### Indian Institute of Technology Roorkee

# CHN-323 Computer Applications in Chemical Engineering

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- ightharpoonup Key idea: Rearrange f(x) = 0 as x = g(x) and then perform iterations as  $x^{[n+1]} = g(x^{[n]})$
- $\triangleright$  Example:  $f(x) = x^2 2x 3 = 0$ 
  - Rearranging:  $x^2 = 2x + 3 \rightarrow x = \sqrt{2x + 3}$
  - Iteration: Let us take x = 4 as initial guess

$$x_0 = 4$$
,  
 $x_1 = \sqrt{11} = 3.31662$ ,  
 $x_2 = \sqrt{9.63325} = 3.10375$ ,  
 $x_3 = \sqrt{9.20750} = 3.03439$ ,  
 $x_4 = \sqrt{9.06877} = 3.01144$ ,  
 $x_5 = \sqrt{9.02288} = 3.00381$ ,

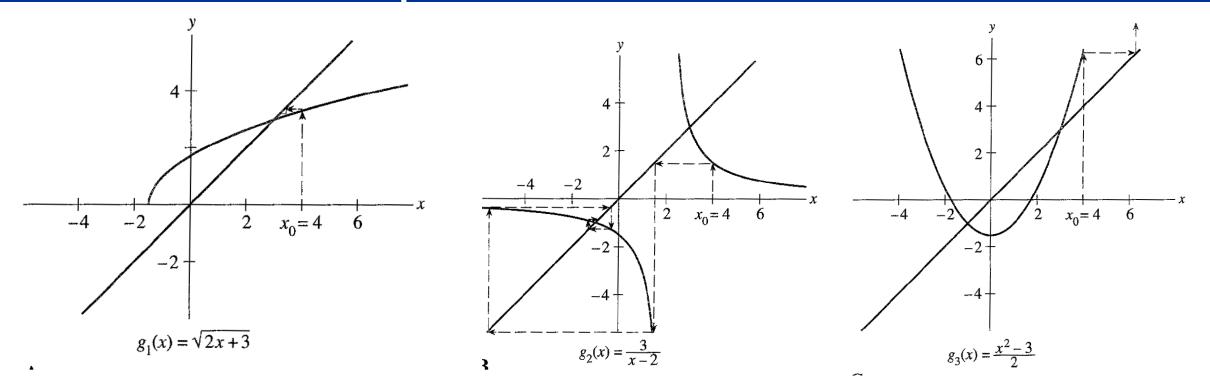
- Rearranging in the form of x = g(x) can be done in multiple ways
- Example:  $f(x) = x^2 2x 3 = 0$ -  $x^2 = 2x + 3 \rightarrow x = \sqrt{2x + 3}$ -  $x = (x^2 - 3)/2$ -  $x(x - 2) = 3 \rightarrow x = 3/(x - 2)$ -  $x^2 = 2x - 3 \rightarrow x = 2 - (3/x)$

- > Let us try another rearrangement
- > Example:  $f(x) = x^2 2x 3 = 0$ 
  - Rearranging:  $x(x 2) = 3 \rightarrow x = 3/(x 2)$
  - Iteration: Let us take x = 4 as initial guess

$$x_0 = 4$$
,  $x_3 = -0.375$ ,  
 $x_1 = 1.5$ ,  $x_4 = -1.263158$ ,  
 $x_2 = -6$ ,  $x_5 = -0.919355$ ,  
 $x_6 = -1.02762$ ,  
 $x_7 = -0.990876$ ,  
 $x_8 = -1.00305$ ,

- > Let us try another rearrangement
- $\triangleright$  Example:  $f(x) = x^2 2x 3 = 0$ 
  - Rearranging:  $x = (x^2 3)/2$
  - Iteration: Let us take x = 4 as initial guess

$$x_0 = 4,$$
  $x_1 = 6.5,$   $x_2 = 19.625,$   $x_3 = 191.070,$  Diverged!



➤ It appears that the different behaviors depend on whether the slope of the curve is greater, less, or of opposite sign to the slope of the line (which equals + 1).

- > Convergence is an issue
  - Not all rearrangements lead to convergence
- > Consider fixed point iteration:

- 
$$x^{[n+1]} = g(x^{[n]})$$

- At convergence, r = g(r)

$$\rightarrow x^{[n+1]} - r = g(x^{[n]}) - g(r)$$

$$\rightarrow e^{[n+1]} = g(x^{[n]}) - g(r)$$

ightharpoonup Multiply and divide RHS with  $x^{[n]}-r$ 

$$- e^{[n+1]} = \left(\frac{g(x^{[n]}) - g(r)}{x^{[n]} - r}\right) \left(x^{[n]} - r\right)$$

> With mean value theorem

$$e^{[n+1]} = g'(\xi)(x^{[n]} - r)$$

where  $\xi$  lies between  $x^{[n]}$  and r

$$\rightarrow e^{[n+1]} = g'(\xi)e^{[n]}$$

- 1. The error will decrease with every iteration if  $|g'(\xi)| < 1$
- 2. The rate of convergence is linear since  $e^{[n+1]} \propto e^{[n]}$

### Error analysis for Newton's method

- $x^{[n+1]} = x^{[n]} \frac{f(x^{[n]})}{f'(x^{[n]})}$
- $\triangleright$  Comparing with the fixed point iteration method (i.e., x = g(x)), we can say

 $g(x) = x - \frac{f(x)}{f'(x)}$ 

 $\succ$  We have found out that the method will converge if  $|g'(\xi)| < 1$ 

$$g'(x) = 1 - \frac{(f')^2 - ff''}{(f')^2} = \frac{ff''}{(f')^2}$$

ightharpoonup At root (x=r), f(r)=0 and  $f'(r)\neq 0$ 

Thus, 
$$g'(r) = 0$$

For Newton scheme, we have |g'(r)| < 1. Therefore, with good initial guess, the Newton Scheme will converge.

### System of nonlinear algebraic equations

$$f_1(x_1, x_2, \dots, x_n) = 0$$
  
 $f_2(x_1, x_2, \dots, x_n) = 0$   
:  
:  
 $f_n(x_1, x_2, \dots, x_n) = 0$ 

- $\triangleright$  In short form,  $f(\underline{x}) = \underline{0}$
- $\succ$  The equations are satisfied at the root ( $\underline{x} = \underline{r}$ ), i.e.,

$$x_1 = r_1, x_2 = r_2, \dots, x_n = r_n$$

> Let us consider two nonlinear equations with two unknowns

$$f_1(x_1, x_2) = 0$$

$$f_2(x_1, x_2) = 0$$

$$\underline{f(\underline{x})} = \underline{0}$$

 $\triangleright$  Assuming that the roots  $\underline{r}$  are  $\{r1, r2\}$ 

Let us revise Taylor series expansion of f(x) around the point  $x^{[0]}$  (for a single variable)

$$f(x) = f(x^{[0]}) + f'(x^{[0]})(x - x^{[0]}) + \frac{f''(x^{[0]})(x - x^{[0]})^2}{2!} + \cdots$$

Taylor series expansion of  $\underline{f}(\underline{x})$  around the point  $(x_1^{[0]}, x_2^{[0]})$  (for two variables)

$$f_{1}(x_{1}, x_{2}) = f_{1}\left(x_{1}^{[0]}, x_{2}^{[0]}\right) + \left[\frac{\partial f_{1}}{\partial x_{1}}\right]_{\left(x_{1}^{[0]}, x_{2}^{[0]}\right)} \left(x_{1} - x_{1}^{[0]}\right) + \left[\frac{\partial f_{1}}{\partial x_{2}}\right]_{\left(x_{1}^{[0]}, x_{2}^{[0]}\right)} \left(x_{2} - x_{2}^{[0]}\right) + \cdots$$

$$f_{2}(x_{1}, x_{2}) = f_{2}\left(x_{1}^{[0]}, x_{2}^{[0]}\right) + \left[\frac{\partial f_{2}}{\partial x_{1}}\right]_{\left(x_{1}^{[0]}, x_{2}^{[0]}\right)} \left(x_{1} - x_{1}^{[0]}\right) + \left[\frac{\partial f_{2}}{\partial x_{2}}\right]_{\left(x_{1}^{[0]}, x_{2}^{[0]}\right)} \left(x_{2} - x_{2}^{[0]}\right) + \cdots$$

 $\succ$  Consider  $\underline{x} = \underline{r}$  (roots)

$$f_{1}(r_{1}, r_{2}) = f_{1}\left(x_{1}^{[0]}, x_{2}^{[0]}\right) + \left[\frac{\partial f_{1}}{\partial x_{1}}\right]_{\left(x_{1}^{[0]}, x_{2}^{[0]}\right)} \left(r_{1} - x_{1}^{[0]}\right) + \left[\frac{\partial f_{1}}{\partial x_{2}}\right]_{\left(x_{1}^{[0]}, x_{2}^{[0]}\right)} \left(r_{2} - x_{2}^{[0]}\right)$$

$$f_{2}(r_{1}, r_{2}) = f_{2}\left(x_{1}^{[0]}, x_{2}^{[0]}\right) + \left[\frac{\partial f_{2}}{\partial x_{1}}\right]_{\left(x_{1}^{[0]}, x_{2}^{[0]}\right)} \left(r_{1} - x_{1}^{[0]}\right) + \left[\frac{\partial f_{2}}{\partial x_{2}}\right]_{\left(x_{1}^{[0]}, x_{2}^{[0]}\right)} \left(r_{2} - x_{2}^{[0]}\right)$$

ightharpoonup But  $\underline{f}(\underline{r}) = \underline{0}$ 

$$0 = f_1 \left( x_1^{[0]}, x_2^{[0]} \right) + \left[ \frac{\partial f_1}{\partial x_1} \right]_{\left( x_1^{[0]}, x_2^{[0]} \right)} \left( r_1 - x_1^{[0]} \right) + \left[ \frac{\partial f_1}{\partial x_2} \right]_{\left( x_1^{[0]}, x_2^{[0]} \right)} \left( r_2 - x_2^{[0]} \right)$$

$$0 = f_2 \left( x_1^{[0]}, x_2^{[0]} \right) + \left[ \frac{\partial f_2}{\partial x_1} \right]_{\left( x_1^{[0]}, x_2^{[0]} \right)} \left( r_1 - x_1^{[0]} \right) + \left[ \frac{\partial f_2}{\partial x_2} \right]_{\left( x_1^{[0]}, x_2^{[0]} \right)} \left( r_2 - x_2^{[0]} \right)$$

 $\succ$  Let us introduce  $\delta_1=r_1-x_1^{[0]}$  and  $\delta_2=r_2-x_2^{[0]}$ 

$$0 = f_1\left(x_1^{[0]}, x_2^{[0]}\right) + \left[\frac{\partial f_1}{\partial x_1}\right]_{\left(x_1^{[0]}, x_2^{[0]}\right)} \delta_1 + \left[\frac{\partial f_1}{\partial x_2}\right]_{\left(x_1^{[0]}, x_2^{[0]}\right)} \delta_2$$

$$0 = f_2\left(x_1^{[0]}, x_2^{[0]}\right) + \left[\frac{\partial f_2}{\partial x_1}\right]_{\left(x_1^{[0]}, x_2^{[0]}\right)} \delta_1 + \left[\frac{\partial f_2}{\partial x_2}\right]_{\left(x_1^{[0]}, x_2^{[0]}\right)} \delta_2$$

> In matrix form

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} f_1 \\ f_2 \end{bmatrix}_{\begin{pmatrix} x_1^{[0]}, x_2^{[0]} \end{pmatrix}} + \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix}_{\begin{pmatrix} x_1^{[0]}, x_2^{[0]} \end{pmatrix}} \begin{bmatrix} \delta_1 \\ \delta_2 \end{bmatrix}$$

> In compact form

$$\underline{0} = \underline{f}^{[0]} + \underline{J}^{[0]}\underline{\delta} \qquad \rightarrow \underline{\delta} = -\left[\underline{\underline{J}}^{[0]}\right]^{-1}\underline{f}^{[0]}$$

The improved guess  $\underline{x}^{[1]}$  is obtained as  $\underline{x}^{[0]} + \underline{\delta}$ . Continue iterations until convergence.

> In general, we can write

$$\underline{\delta}^{[n]} = -\left[\underline{\underline{J}}^{[n]}\right]^{-1} \underline{f}^{[n]}$$

$$\underline{x}^{[n+1]} = \underline{x}^{[n]} + \underline{\delta}^{[n]}$$

#### Example

> 5 CSTR Of equal volume operating in series at steady state. Reaction  $(A \rightarrow B)$  in CSTR is second order, reaction rate constant is unity. Volumetric flow rate at inlet of 1<sup>st</sup> reactor is 200 m<sup>3</sup>/s. In the figure below, x denotes concentration of A. Find Volume of each reactor.

