

18-660: Numerical Methods for Engineering Design and Optimization

Xin Li

Department of ECE

Carnegie Mellon University

Pittsburgh, PA 15213

Overview

- Nonlinear Equation Solver
 - ▼ Newton-Raphson method
 - ▼ Binary search

Nonlinear Algebraic Equation

- Many physical systems are nonlinear and must be mathematically described as a nonlinear equation

$$F(x) = 0$$

- Example: nonlinear ordinary differential equation

$$F[\dot{x}(t), x(t), u(t)] = 0 \quad x(0) = X$$

$x(t)$: N-dimensional vector of unknown variables

$u(t)$: Vector of input sources

F : Nonlinear operator

X : Initial condition

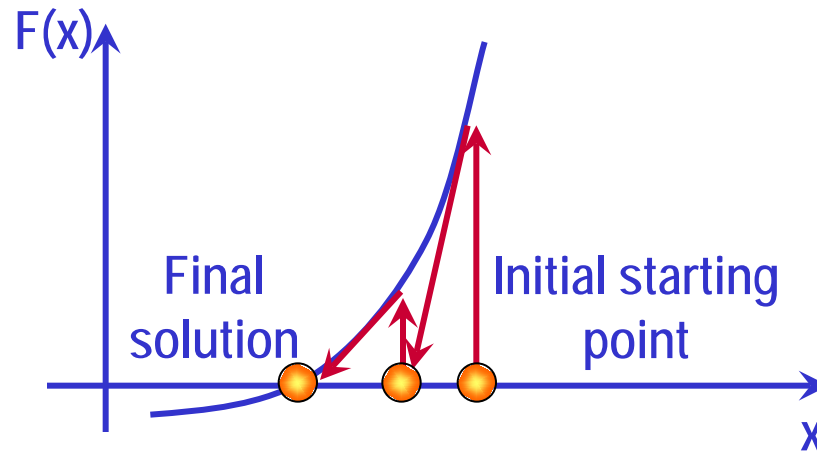
Nonlinear Algebraic Equation

$$F(x) = 0$$

- Closed-form solution cannot be derived for a general nonlinear equation
- Iterative algorithm must be applied to find an approximate solution (i.e., numerical solution)
- Newton-Raphson method is a widely-used algorithm to solve nonlinear algebraic equation

Newton-Raphson Method

■ A one-dimensional example



$$F(x) = 0 \quad \Rightarrow \quad \left\{ \begin{array}{l} \text{k-th iteration:} \quad F[x^{(k)}] \neq 0 \\ \text{Update solution:} \quad x^{(k+1)} = x^{(k)} + \Delta x^{(k)} \end{array} \right.$$

Linearize $F(x)$ via first-order Taylor series expansion
and solve iteratively

Newton-Raphson Method

- Taylor series approximation

- ▼ For a convergent series and sufficiently small $\Delta x^{(k)}$

$$F[x^{(k)} + \Delta x^{(k)}] = F[x^{(k)}] + F'[x^{(k)}] \cdot \Delta x^{(k)} + \dots$$

- Newton-Raphson method relies on 1st-order Taylor expansion

- ▼ The function $F(\bullet)$ must be smooth and its derivative exists

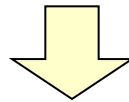
Newton-Raphson Method

- Apply first-order Taylor expansion

$$F[x^{(k+1)}] \approx F[x^{(k)}] + F'[x^{(k)}] \cdot \underbrace{[x^{(k+1)} - x^{(k)}]}_{\Delta x^{(k)}} = 0$$



$$\Delta x^{(k)} = -\frac{F[x^{(k)}]}{F'[x^{(k)}]}$$

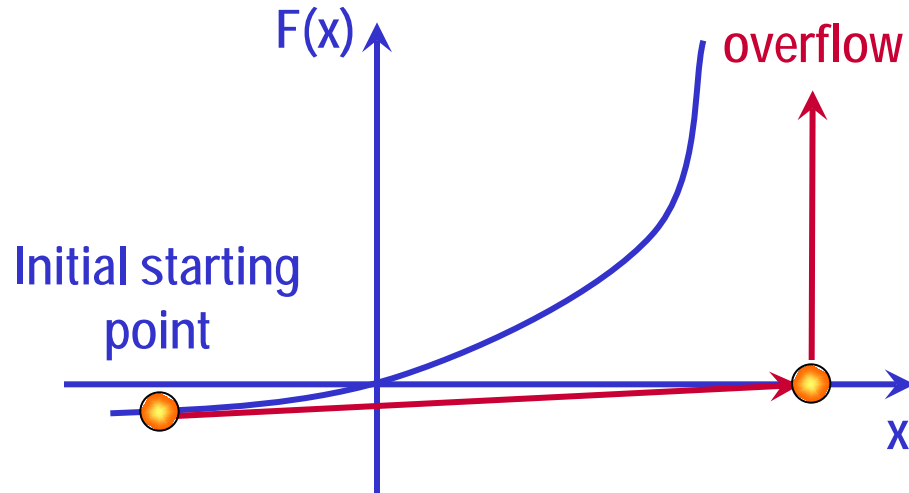


$$x^{(k+1)} = x^{(k)} - \frac{F[x^{(k)}]}{F'[x^{(k)}]}$$

- Global convergence is not guaranteed in general
 - ▼ Many techniques exist to improve convergence

Newton-Raphson Method

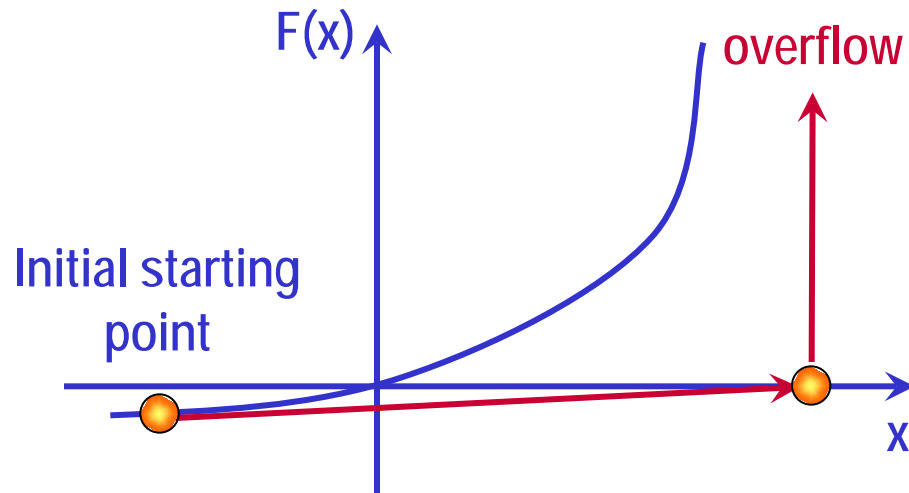
- A simple example:



$$x^{(k+1)} = x^{(k)} - \frac{F[x^{(k)}]}{F'[x^{(k)}]}$$

Newton-Raphson Method

- A simple example:



$$x^{(k+1)} = x^{(k)} - \lambda \cdot \frac{F[x^{(k)}]}{F'[x^{(k)}]} \quad (0 < \lambda < 1)$$

Damping can be applied to limit the extent of the linearized projection within one iteration

Newton-Raphson Method

$$F[x^{(k+1)}] \approx F[x^{(k)}] + F'[x^{(k)}] \cdot [x^{(k+1)} - x^{(k)}] = 0$$

Accurate if $x^{(k)}$ and $x^{(k+1)}$ are close

- Start Newton-Raphson iteration from a good initial solution
- In practice, a good initial solution may be unknown
- An alternative approach is to randomly select multiple initial solutions and apply Newton-Raphson method to each of them

A Simple Example

$$F(x) = \log(x) = 0$$

$$F'(x) = \frac{1}{x}$$

- Start from an initial solution $x^{(0)} = 0.1$

$$F[x^{(0)}] = \log(0.1) = -2.30$$

$$F'[x^{(0)}] = \frac{1}{0.1} = 10$$

A Simple Example

$$F(x) = \log(x) = 0$$

$$F'(x) = \frac{1}{x}$$

$$x^{(0)} = 0.1$$

$$F[x^{(0)}] = -2.30$$

$$F'[x^{(0)}] = 10$$

■ Iteration #1

$$F[x^{(1)}] \approx F[x^{(0)}] + F'[x^{(0)}] \cdot [x^{(1)} - x^{(0)}] = 0$$

$$-2.3 + 10 \cdot [x^{(1)} - 0.1] = 0$$

$$x^{(1)} = 0.33$$

$$F[x^{(1)}] = \log(0.33) = -1.11$$

$$F'[x^{(1)}] = \frac{1}{0.33} = 3$$

A Simple Example

$$F(x) = \log(x) = 0$$

$$F'(x) = \frac{1}{x}$$

$$x^{(1)} = 0.33$$

$$F[x^{(1)}] = -1.11$$

$$F'[x^{(1)}] = 3$$

■ Iteration #2

$$F[x^{(2)}] \approx F[x^{(1)}] + F'[x^{(1)}] \cdot [x^{(2)} - x^{(1)}] = 0$$

$$-1.11 + 3 \cdot [x^{(2)} - 0.33] = 0$$

$$x^{(2)} = 0.70$$

$$F[x^{(2)}] = \log(0.70) = -0.36$$

$$F'[x^{(2)}] = \frac{1}{0.70} = 1.43$$

A Simple Example

$$F(x) = \log(x) = 0$$

$$F'(x) = \frac{1}{x}$$

$$x^{(2)} = 0.70$$

$$F[x^{(2)}] = -0.36$$

$$F'[x^{(2)}] = 1.43$$

■ Iteration #3

$$F[x^{(3)}] \approx F[x^{(2)}] + F'[x^{(2)}] \cdot [x^{(3)} - x^{(2)}] = 0$$

$$-0.36 + 1.43 \cdot [x^{(3)} - 0.70] = 0$$

$$x^{(3)} = 0.95$$

$$F[x^{(3)}] = \log(0.95) = -0.05$$

$$F'[x^{(3)}] = \frac{1}{0.95} = 1.05$$

Newton-Raphson Method

■ Extend to multi-dimensional case

$$F(x) = 0 \quad \begin{cases} F_1(x) = 0 \\ F_2(x) = 0 \\ \vdots \end{cases}$$

▼ Apply first-order Taylor expansion for all F_i 's

$$F_1[x^{(k+1)}] = F_1[x^{(k)}] + \left. \frac{\partial F_1}{\partial x_1} \right|_k \cdot [x_1^{(k+1)} - x_1^{(k)}] + \left. \frac{\partial F_1}{\partial x_2} \right|_k \cdot [x_2^{(k+1)} - x_2^{(k)}] + \dots = 0$$

$$F_2[x^{(k+1)}] = F_2[x^{(k)}] + \left. \frac{\partial F_2}{\partial x_1} \right|_k \cdot [x_1^{(k+1)} - x_1^{(k)}] + \left. \frac{\partial F_2}{\partial x_2} \right|_k \cdot [x_2^{(k+1)} - x_2^{(k)}] + \dots = 0$$

⋮

Newton-Raphson Method

- Re-write into a matrix form

$$\begin{bmatrix} F_1[x^{(k+1)}] \\ F_2[x^{(k+1)}] \\ \vdots \end{bmatrix} = \begin{bmatrix} F_1[x^{(k)}] \\ F_2[x^{(k)}] \\ \vdots \end{bmatrix} + \begin{bmatrix} \left. \frac{\partial F_1}{\partial x_1} \right|_k & \left. \frac{\partial F_1}{\partial x_2} \right|_k & \dots \\ \left. \frac{\partial F_2}{\partial x_1} \right|_k & \left. \frac{\partial F_2}{\partial x_2} \right|_k & \dots \\ \vdots & \vdots & \vdots \end{bmatrix} \cdot \left(\begin{bmatrix} x_1^{(k+1)} \\ x_2^{(k+1)} \\ \vdots \end{bmatrix} - \begin{bmatrix} x_1^{(k)} \\ x_2^{(k)} \\ \vdots \end{bmatrix} \right) = 0$$

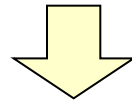
$\frac{\partial F}{\partial x} \Big|_k$ **Jacobian matrix**

$\Delta x^{(k)}$

Newton-Raphson Method

- The new $\Delta x^{(k)}$ can be solved by a linear solver

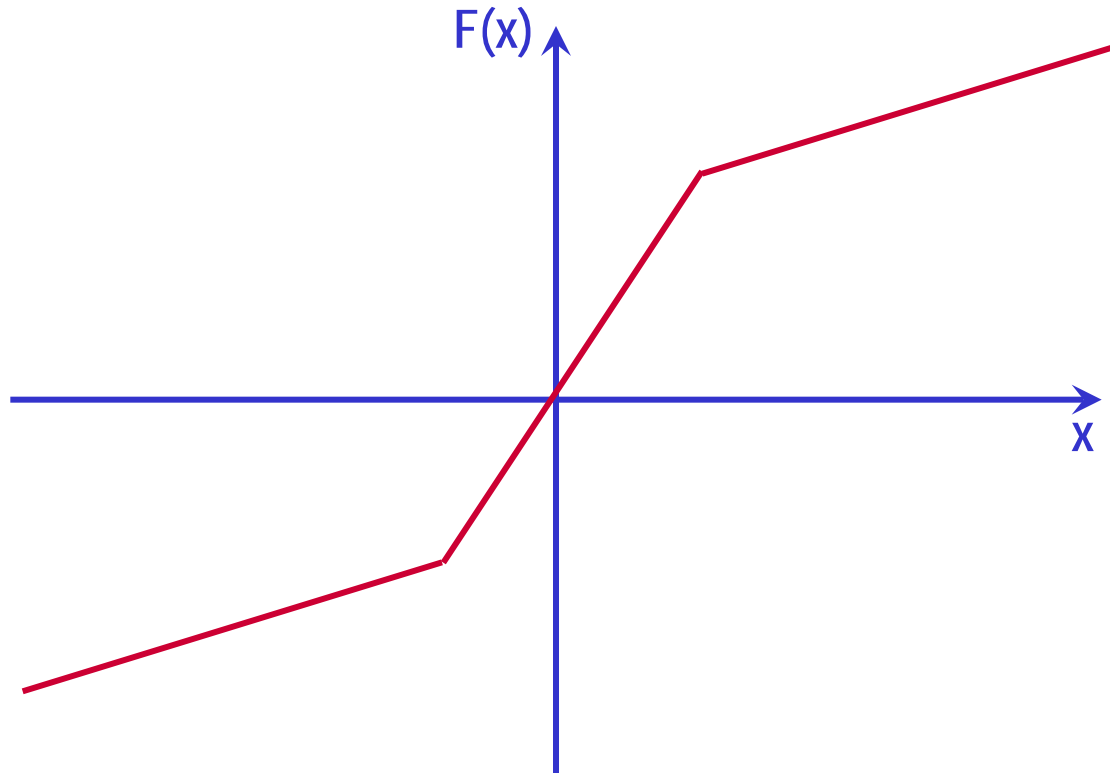
$$F[x^{(k+1)}] \approx F[x^{(k)}] + \left. \frac{\partial F}{\partial x} \right|_k \cdot \Delta x^{(k)} = 0$$



$$\underbrace{\Delta x^{(k)}}_{\text{Vector}} = - \underbrace{\left[\left. \frac{\partial F}{\partial x} \right|_k \right]^{-1}}_{\text{Matrix}} \cdot \underbrace{F[x^{(k)}]}_{\text{Vector}}$$

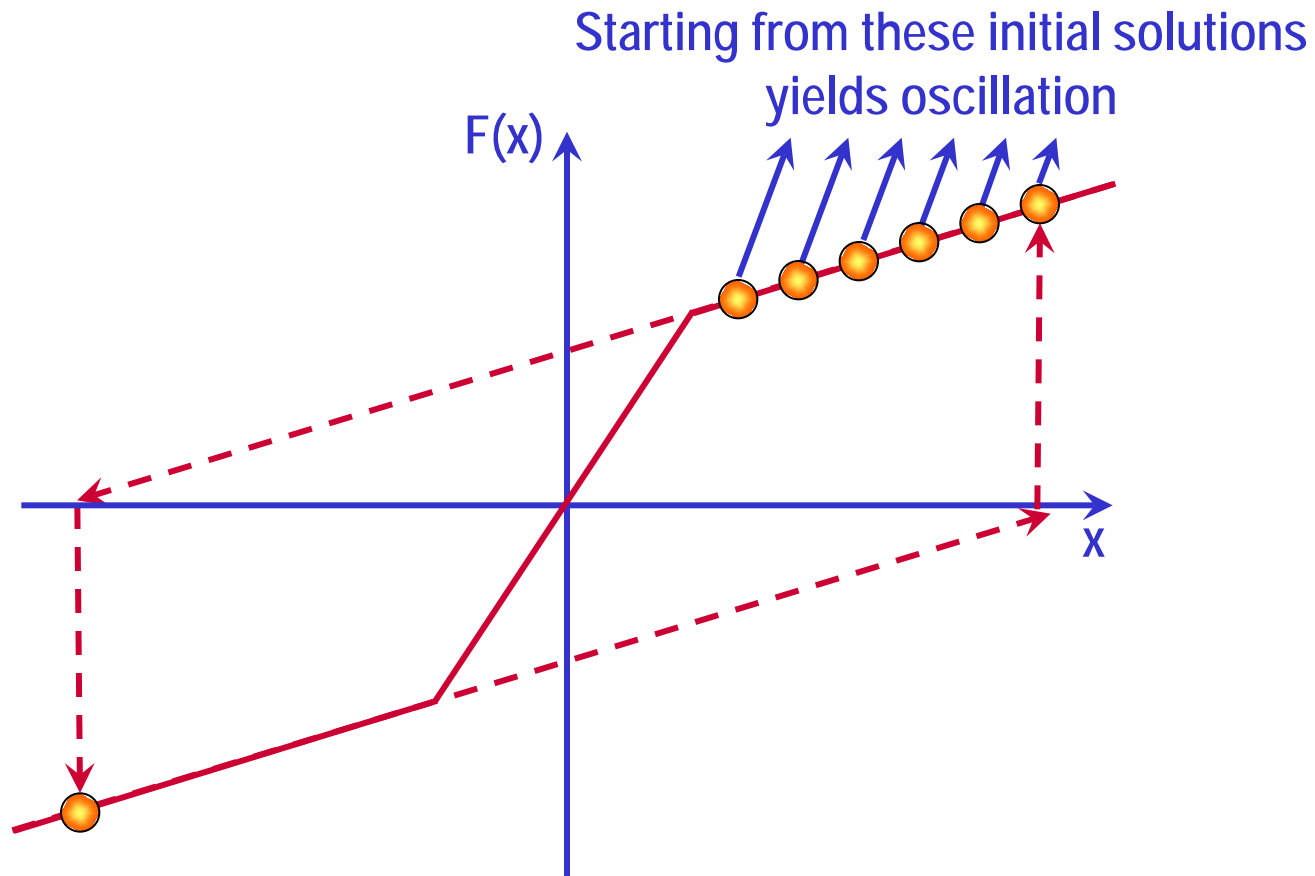
Newton-Raphson Method

- Newton-Raphson method is applicable, if and only if $F(x)$ is smooth and its derivative exists
- A bad example: piecewise-linear function



Newton-Raphson Method

- If we use Newton-Raphson to solve the problem
 - ▼ 2nd order derivative is not continuous



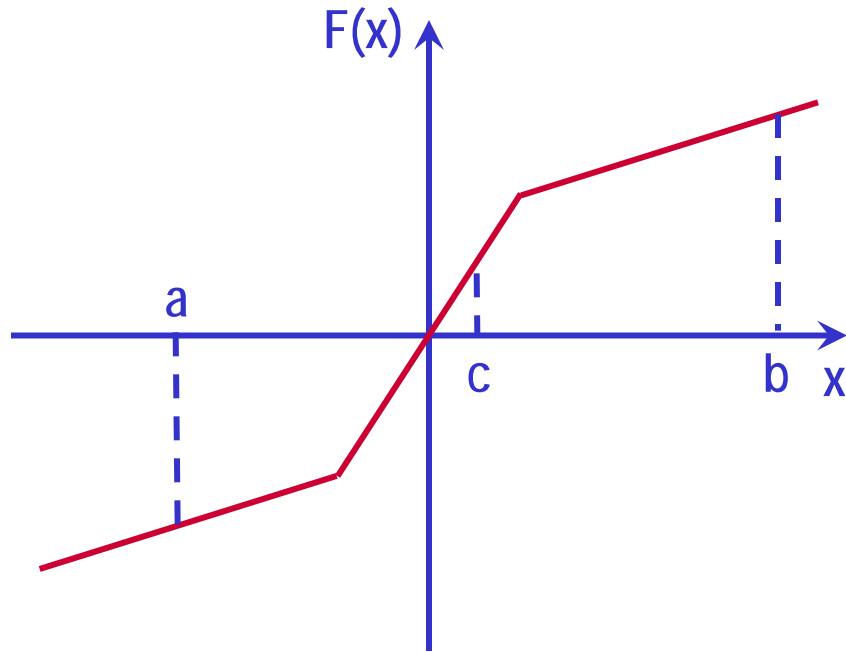
Binary Search

- Binary search is an alternative algorithm that is efficient for **one-dimensional** non-smooth but **continuous** function
- Key idea
 - ▼ Assume that the solution of $F(x) = 0$ is within an interval $[a, b]$
 - ▼ Iteratively shrink $[a, b]$ to find the solution x

Binary Search

■ Select initial interval $[a, b]$ $F(a) \cdot F(b) < 0$

■ Set c as the center of $[a, b]$ $c = \frac{a+b}{2}$

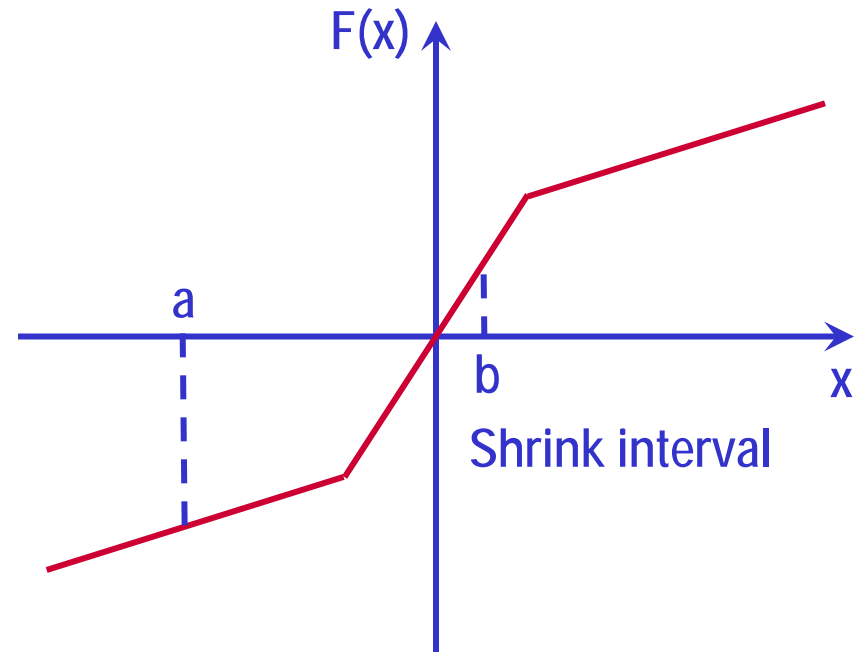
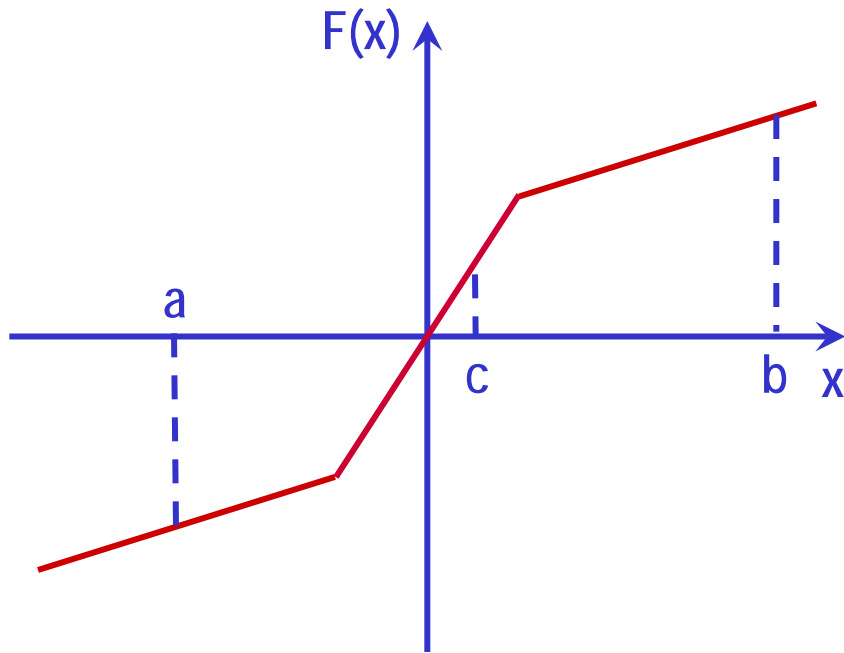


Binary Search

■ If $F(a) \cdot F(c) < 0$

■ Then $b = c$

Solution is within $[a, c]$



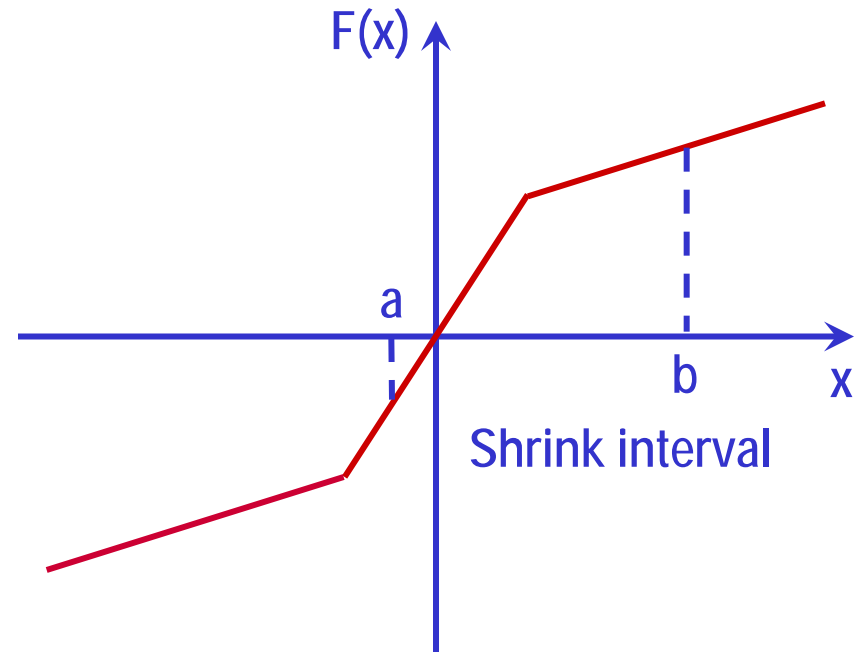
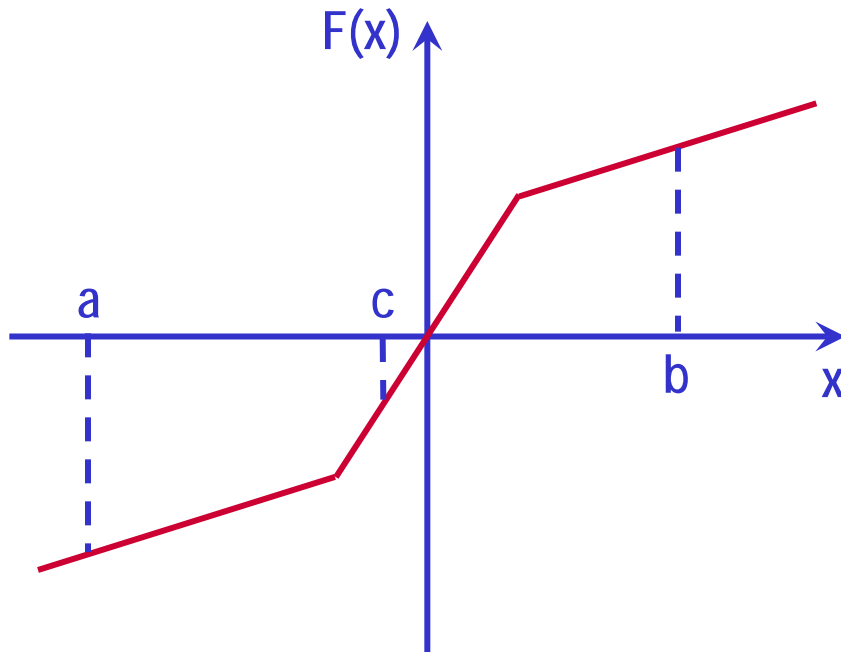
Binary Search

■ If $F(b) \cdot F(c) < 0$

■ Then $a = c$

Solution is within $[c, b]$

Continue iteration until convergence is reached



A Simple Example

$$\log(x) = 0 \quad \text{where} \quad x \in [0, 10]$$

■ Iteration #1

$$\begin{array}{ll} a = 0 & \log(a) < 0 \\ b = 10 & \log(b) > 0 \\ c = 5 & \log(c) > 0 \end{array} \quad \Rightarrow \quad x \in [0, 5]$$

■ Iteration #2

$$\begin{array}{ll} a = 0 & \log(a) < 0 \\ b = 5 & \log(b) > 0 \\ c = 2.5 & \log(c) > 0 \end{array} \quad \Rightarrow \quad x \in [0, 2.5]$$

A Simple Example

$$\log(x) = 0 \quad \text{where} \quad x \in [0, 10]$$

■ Iteration #3

$$\begin{array}{ll} a = 0 & \log(a) < 0 \\ b = 2.5 & \log(b) > 0 \\ c = 1.25 & \log(c) > 0 \end{array} \quad \Rightarrow \quad x \in [0, 1.25]$$

■ Iteration #4

$$\begin{array}{ll} a = 0 & \log(a) < 0 \\ b = 1.25 & \log(b) > 0 \\ c = 0.625 & \log(c) < 0 \end{array} \quad \Rightarrow \quad x \in [0.625, 1.25]$$

Continue iteration until convergence is reached

Summary

- Nonlinear equation solver
 - ▼ Newton-Raphson method
 - ▼ Binary search