

CS 450: Numerical Analysis

Lecture 14

Chapter 5 – Nonlinear Equations

Solving Systems of Nonlinear Equations

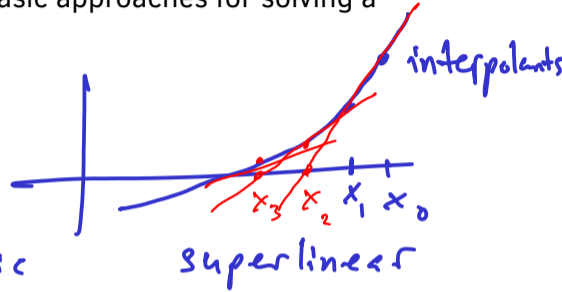
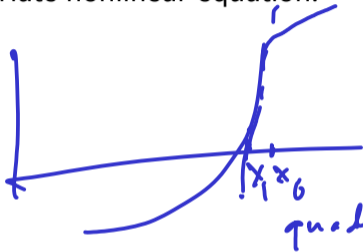
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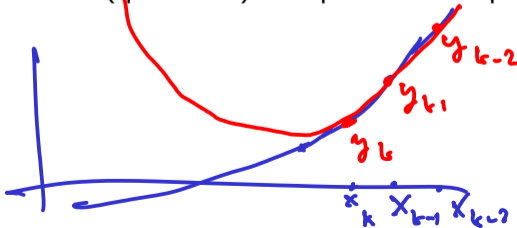
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Review Solving a Nonlinear Equation

- ▶ Newton's and secant method provide basic approaches for solving a univariate nonlinear equation:



- ▶ Inverse (quadratic) interpolation can provide better convergence:



$$\begin{matrix}
 \left(\begin{array}{cc}
 x_{k-2} & y_{k-2} \\
 x_{k-1} & y_{k-1} \\
 x_k & y_k
 \end{array} \right) \\
 y = p(x) \\
 \boxed{x = p(y)} \\
 \begin{matrix}
 x_{k+1} \\
 p^{-1} \\
 p(0)
 \end{matrix}
 \end{matrix}$$

Systems of Nonlinear Equations

- Given $f(x) = \begin{bmatrix} f_1(x) \\ \vdots \\ f_m(x) \end{bmatrix}$ for $x \in \mathbb{R}^n$, seek $x^* \in \mathbb{R}^n$ so that $f(x^*) = 0$

$$f(x^*) = \begin{bmatrix} f_1(x^*) \\ \vdots \\ f_m(x^*) \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

- At a particular point x , the *Jacobian* of f , describes how f changes in a given direction of change in x ,

$$J_f(x) = \begin{bmatrix} \frac{df_1}{dx_1}(x) & \cdots & \frac{df_1}{dx_n}(x) \\ \vdots & & \vdots \\ \frac{df_m}{dx_1}(x) & \cdots & \frac{df_m}{dx_n}(x) \end{bmatrix}$$

Newton's method

$$\hat{f}(x + \delta x) = 0$$

$$f(x) = -J_f(x) \delta x$$

$$\delta x = -J_f^{-1}(x) f(x) \quad f(x + \delta x) \approx f(x) + J_f(x) \delta x = \hat{f}(x)$$

Multivariate Fixed-Point and Newton Iteration

- ▶ Fixed-point iteration $\mathbf{x}_{k+1} = \mathbf{g}(\mathbf{x}_k)$ achieves local convergence so long as $|\lambda_{\max}(\mathbf{J}_g(\mathbf{x}^*))| < 1$:

$$\text{Spectral radius}(A) = \frac{|\lambda_{\max}(A)|}{\rho(A)}$$

- ▶ Newton's method corresponds to the fixed-point iteration

$$\mathbf{g}(\mathbf{x}) = \mathbf{x} - \mathbf{J}_f^{-1}(\mathbf{x})\mathbf{f}(\mathbf{x})$$

$$\mathbf{g}(\mathbf{x}^*) = \mathbf{x}^*$$

$$\mathbf{g}(\mathbf{x}^*) = \mathbf{x}^* \quad \text{whenever } \mathbf{f}(\mathbf{x}^*) = \mathbf{0}$$

$$\mathbf{J}_f(\mathbf{x})\mathbf{s} = \mathbf{f}(\mathbf{x}) \quad \Bigg| \quad \mathbf{J}_f(\mathbf{x}^*)\mathbf{s} = \mathbf{f}(\mathbf{x}^*) = \mathbf{0}$$

Convergence of Newton Iteration

- ▶ Newton's method achieves quadratic local convergence if $\|J_f^{-1}(x^*)\|$ is bounded:

$$e_k = x_k - x^* = g(x_{k-1}) - x^*$$

$$= x_{k-1} - J_f^{-1}(x_{k-1}) f(x_{k-1}) - x^*$$

$$= -J_f^{-1}(x_{k-1}) \left[\underbrace{f(x_{k-1}) + J_f(x_{k-1}) [x_{k-1} - x^*]}_{f(x^*) - (f(x_{k-1}) + J_f(x_{k-1}) [x_{k-1} - x^*])} \right]$$

$$\|e_k\| \leq \|J_f^{-1}(x_{k-1})\| \underbrace{\|f(x^*) - (f(x_{k-1}) + J_f(x_{k-1}) [x_{k-1} - x^*])\|}_{\|x_{k-1} - x^*\|^2} = \mathcal{O}(\|e_{k-1}\|^2)$$

Convergence of Newton Iteration (II)

- ▶ Quadratic convergence is achieved when the Jacobian of a fixed-point iteration is zero at the solution:

$$g(x) = x - \underline{J_f(x)^{-1}} f(x) \quad \underline{J_g(x^*)} = 0$$

$$J_g(x) = I - \underbrace{J_f(x)^{-1} J_f(x)}_I - f(x) \cdot \frac{d J_f(x)}{dx}$$

$$J_g(x^*) = \underbrace{I - I}_{0} - \underbrace{f(x^*)}_{0} \cdot \frac{d J_f(x)}{dx}$$

$$\underbrace{f(x^*)}_{0} = 0$$

Estimating the Jacobian using Finite Differences

- ▶ To obtain $J_f(x_k)$ at iteration k , can use finite differences:

$$n=1 \quad x_k \in \mathbb{R}$$

$$f(x) \in \mathbb{R}^m$$

$$f(x) = \begin{bmatrix} f_1(x) \\ \vdots \\ f_n(x) \end{bmatrix}$$

$$J_f(x_k) \approx \frac{f(x+h) - f(x)}{h}$$

$$j_i \text{ as } i\text{th column of } J_f(x_k)$$

- ▶ How many function evaluations are generally needed?

evaluate

$$\left. \begin{array}{l} f(x) \\ f(x+h e_1) \\ \vdots \\ f(x+h e_n) \end{array} \right\} n+1$$

$$j_i \approx \frac{f(x+h e_i) - f(x)}{h}$$