

# Midterm Review

**Secant Method:**  $x_{k+1} = x_k - f(x_k) \frac{x_k - x_{k-1}}{f(x_k) - f(x_{k-1})}$

**Newton's method:**  $x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}$

rate of convergence:  $\left| \frac{f''(\eta)}{2f'(x_k)} \right| \rightarrow \left| \frac{f''(\xi)}{2f'(\xi)} \right|$  convergence  $\leftrightarrow \left| \frac{f''(x)}{f'(y)} \right| \leq A$

$$0 = f(\xi) = f(x_k) + f'(x_k)(\xi - x_k) + \frac{f''(\eta)}{2}(\xi - x_k)^2$$

$$\xi - x_k = -\frac{(\xi - x_k)^2 f''(\eta)}{2f'(x_k)} - \frac{f(x_k)}{f'(x_k)} \quad \xi - x_{k+1} = -\frac{(\xi - x_k)^2 f''(\eta)}{2f'(x_k)}$$

$$\lim_{k \rightarrow \infty} \frac{|x_{k+1} - \xi|}{|x_k - \xi|^2} = \lim_{k \rightarrow \infty} \left| \frac{f''(\eta)}{2f'(x_k)} \right| = \left| \frac{f''(\xi)}{2f'(\xi)} \right|$$

**Bisection method:**  $c_k = \frac{1}{2}(a_k + b_k)$

$$(a_{k+1}, b_{k+1}) = (a_k, c_k) \text{ if } f(c_k)f(b_k) > 0$$

$$= (c_k, b_k) \text{ if } f(c_k)f(b_k) < 0$$

A vector norm on  $X$  is a function  $p : X \rightarrow R$

$$1. p(x + y) \leq p(x) + p(y)$$

$$2. p(sx) = |s|p(x)$$

$$3. \text{ for all } x \in X, p(x) = 0 \rightarrow x = 0 \quad \text{or} \quad p(x) = 0 \leftrightarrow x = 0$$

$$\|v\|_1 = |v_1| + \dots + |v_n|$$

$$\|v\|_2 = \sqrt{v_1^2 + \dots + v_n^2}$$

$$\|v\|_\infty = \max |v_i|$$

$$\|v\|_p = (|v_1|^p + \dots + |v_n|^p)^{\frac{1}{p}}$$

A matrix norm is a vector norm in a vector space whose elements are matrices, and  $\|AB\| \leq \|A\| \|B\|$

$$\|A\|_p = \max \frac{\|Ax\|_p}{\|x\|_p}$$

$\|A\|_1$  is the maximum absolute column sum of the matrix

$\|A\|_\infty$  is the maximum absolute row sum of the matrix

$\|A\|_2 = \sqrt{\lambda_{\max}(A^T A)} = \sigma_{\max}(A^T A)$  is the largest singular value (squareroot of eigenvalue)

$$Ax = b$$

$$LUx = b \quad Ly = b \quad \text{ForwardSolve} \rightarrow y \quad Ux = y \quad \text{BackwardSolve} \rightarrow x$$

$$QRx = b \quad x = Q^T R^{-1}b \quad \text{Gram Schmidt} \rightarrow Q \quad R = Q^T A$$