Standard vs. Nested Form

Methods

Bisection Method

Standard Form: $a_0 + a_1x + a_2x^2 + \cdots + a_nx^n$ Nested Form: $a_0 + x(a_1 + a_2x + \cdots + a_nx^{n-1})$

Sig Fig Rounding

4-Digit Example	
37.21429	37.21
0.002271828	0.002272
3000527.11059	3001000

Bit Preciscion

	sign	exponent	mantissa	macnine ϵ
single	1	8	23	2^{-23}
double (default)	1	11	52	2^{-52}
long double	1	15	64	2^{-64}
				D

sign mantissa exponent	t
------------------------	---

$$\rightarrow 1.b_1b_2\cdots b_n\times 2^P$$

Error Calculation

$$x_e = \text{Exact}, \ x_a = \text{Approximate}$$
Abs err = $|x_e - x_a|$
Rel err = $\frac{|x_e - x_a|}{x_e}$

Worst case error = $0.5 \cdot 10^{-P+1}$

Loss of Significance

$$a - b = \frac{a^2 - b^2}{a + b}$$

Theorems

Intermediate Value Theorem

If f(x) is continuous on [a, b] and $f(a) \neq f(b)$, then for any y in $[f(a), f(b)] \exists c \in [a, b]$ such that f(c) = y

Mean Value Theorem

If
$$f'(x)$$
 is continuous on $[a, b]$,
then $\exists c \in [a, b]$ such that $f'(c) = (f(b) - f(a))/(b - a)$

Rolle's Theorem

If
$$f'(x)$$
 is continuous on $[a, b]$ and $f(a) = f(b)$,
then $\exists c \in [a, b]$ such that $f'(c) = 0$

Taylor Series

$$P_k(x) = \sum_{n=0}^k \frac{f^{(n)}(a)}{n!} (x - a)^n$$
$$E_{k+1} = \frac{f^{(k+1)}(c)}{(k+1)!} (x - x_0)^{k+1}$$

1. Choose a, b such that $f(a) \cdot f(b) < 0$

- 2. Compute $c = \frac{a+b}{2}$
- 3. If f(c) and f(a) have opposite signs, then b=c
- 4. If f(c) and f(b) have opposite signs, then a=c
- 5. If $|b-a| < \epsilon$ or f(c) = 0 then c is the solution

$$\text{Error} = \frac{b-a}{2^{n+1}}$$

$$\text{Steps}(\epsilon) = \left\lceil \frac{\ln(b-a) - \ln(\epsilon)}{\ln(2)} - 1 \right\rceil$$

Newton's Method

1.
$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

2. If
$$|x_1 - x_0| < \epsilon$$
 or $f(x_1) = 0$ then x_1 is the solution

Multiplicity

If
$$f^{(m-1)}(r_0) = 0$$
 and $f^{(m)}(r_0) \neq 0$,
then $f^{(m)}(r_0)$ is a root of multiplicity m

Convergence
$$m=1$$
 $m>1$ $M=\frac{1}{2}\left|\frac{f''(r_0)}{f'(r_0)}\right|$ $e_{i+1}\approx \frac{m-1}{m}e_i$ $e_{i+1}\approx Me_i^2$

Secant Method

$$x_{n+1} = x_n - f(x_n) \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})}$$

Convergence

$$e_{i+1} \approx M e_i^p$$

$$p = \frac{1+\sqrt{5}}{2} \text{ (Golden Ratio)}$$

False Position Method

Same as Bisection Method, but

$$c = a - \frac{f(a)(b-a)}{f(b) - f(a)}$$

Convergence - Generally same as Bisection **Fixed Point Iteration**

1. Solve f(x) for x = g(x)

$$g(x) = 2.8x - x^{2} = x$$
$$1.8x - x^{2} = 0$$
$$x(1.8 - x) = 0$$
$$x = 0.1.8$$

$$f(x) = 2x^3 - 6x - 1$$
$$g(x) = \frac{2x^3 - 1}{6}$$

2. If $|g'(r_n)| < 1$, fixed point will converge to r_n

Matrix

Gaussian Elimination $O(\frac{2n^3}{3})$

Back Substitution
$$O(n^2)$$

Summations

$$\sum_{i=1}^{n} i = \frac{n(n+1)}{2} \qquad \sum_{i=1}^{n+5} (k-1) = \sum_{i=1}^{n} i^2 = \frac{n(n+1)(2n+1)}{6}$$

$$\sum_{i=1}^{n} i = \frac{n(n+1)}{2} \qquad \sum_{i=1}^{n+5} (k-1) = \sum_{i=1}^{n+4} j = \frac{(n+4)(n+5)}{2} \qquad \frac{s_b^2}{2 \cdot s_e^3/3} = \frac{t_b}{t_e}$$

$$\frac{s_b^2}{2 \cdot s_e^3/3} = \frac{t_b}{t_e}$$

Complexity Ratio

LU Factorization

Goal: $A = LU \to A\overrightarrow{x} = \overrightarrow{b}$

Slow Method

- 1. Row reduce A, recording operations as elementary matrices
- 2. The row reduced A is now U
- 3. You now have $E_n E_{n-1} \cdots E_2 E_1 A = LU$
- 4. Invert the elementary matricies to find L
- 5. $L = (E_n E_{n-1} \cdots E_2 E_1)^{-1}$

$$A = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & -2 \\ -3 & 1 & 1 \end{bmatrix}$$
$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -\frac{7}{3} & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 3 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Example

$$E_3E_2E_1A = U$$

$$A = E_1^{-1}E_2^{-1}E_3^{-1}U$$

$$= (E_3E_2E_1)^{-1}U$$

$$A = LU = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -3 & -\frac{7}{3} & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & -1 \\ 0 & -3 & 0 \\ 0 & 0 & -2 \end{bmatrix}$$

Fast Method

- 1. Apply pivots internally within A
- 2. Note that you will subtract rather than add with these operations

$$\begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & -2 \\ -3 & 1 & 1 \end{bmatrix}$$
 (1)
$$\begin{bmatrix} 1 & 2 & -1 \\ \frac{2}{1} & -3 & 0 \\ -\frac{3}{1} & -7 & -2 \end{bmatrix}$$
 (2)

3. Don't forget to recurse through the inner matrix (row 2 and beyond)

$$\begin{bmatrix} 1 & 2 & -1 \\ \frac{2}{1} & -3 & 0 \\ -\frac{3}{1} & \frac{7}{3} & -2 \end{bmatrix}$$
 (3)

- 4. The circled (lower triangular) elements are now L (plus the identity matrix)
- 5. The remaining upper triangular elements are U

PA = LU Factorization

Goal: $PA = LU \rightarrow A\overrightarrow{x} = \overrightarrow{b}$

- 1. Same logic as LU, but move the |biggest| element to the top of the column
- 2. Make sure you do the same for inner matricies
- 3. You'll have multiple P matrices, such that

$$PA = LU$$

4. Make sure you apply P when solving for \overrightarrow{x}

Solving

- 1. Solve $L\overrightarrow{y} = \overrightarrow{b}$ for \overrightarrow{y}
- 2. Solve $U\overrightarrow{x} = \overrightarrow{y}$ for \overrightarrow{x}