

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	machine ϵ
single	1	8	23	2^{-23}
double (default)	1	11	52	2^{-52}
long double	1	15	64	2^{-64}

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

Error Calculation

x_e = Exact, x_a = 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

$$P_k(x) = \sum_{n=0}^k \frac{f^{(n)}(a)}{n!} (x - a)^n$$

$$E_{k+1} = \frac{f^{(k+1)}(c)}{(k+1)!} (x - x_0)^{k+1}$$

Bisection Method

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

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

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

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

$m > 1$

$$M = \frac{1}{2} \left| \frac{f''(r_0)}{f'(r_0)} \right|$$

$$e_{i+1} \approx \frac{m-1}{m} e_i$$

$$e_{i+1} \approx M e_i^2$$

Secant Method

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

Convergence

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

$$p = \frac{1 + \sqrt{5}}{2}$$
 (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

$O(n^2)$

LU Fact.

Gaussian + 2 * Back Sub

Summations

$$\sum_{i=1}^n i = \frac{n(n+1)}{2}$$

$$\sum_{i=1}^{n+5} (k-1) = \sum_{i=1}^{n+4} j = \frac{(n+4)(n+5)}{2}$$

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

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

LU Factorization

Goal: $A = LU$

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 matrices to find L
- 5. $L = (E_n E_{n-1} \cdots E_2 E_1)^{-1}$

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

$$\begin{matrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -\frac{7}{3} & 1 \end{bmatrix} & \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 3 & 0 & 1 \end{bmatrix} & \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \\ e_3 & e_2 & e_1 \end{matrix}$$

Example

$$E_3 E_2 E_1 A = U$$

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

$$= (E_3 E_2 E_1)^{-1} U$$

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

Fast Method

- 1. Do that shit in one step