

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 $\epsilon$
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

$$P_k(x) = \sum_{n=0}^k \frac{f^{(n)}(x_0)}{n!} (x - x_0)^n$$

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

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

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

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

Newton's Method

1.  $x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$
2. If  $|x_n - x_{n-1}| < \epsilon$  or  $f(x_n) = 0$  then  $x_n$  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

$m > 1$

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

$$e_{i+1} \approx Me_i^2$$

Secant Method

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

Convergence

$$e_{i+1} \approx Me_i^p$$

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

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	Summations	Complexity Ratio
<b>Gaussian Elimination</b> $O(\frac{2n^3}{3})$	$\sum_{i=1}^n i = \frac{n(n+1)}{2}$	$\sum_{i=1}^{n+5} (k-1) = \sum_{i=1}^{n+4} j = \frac{(n+4)(n+5)}{2}$
<b>Back Substitution</b> $O(n^2)$	$\sum_{i=1}^n i^2 = \frac{n(n+1)(2n+1)}{6}$	$\frac{s_b^2}{2 \cdot s_e^3/3} = \frac{t_b}{t_e}$
<b>LU Fact.</b> Gaussian + 2 * Back Sub		

## LU Factorization

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

### 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 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

$$\begin{aligned} E_3 E_2 E_1 A &= U \\ A &= E_1^{-1} E_2^{-1} E_3^{-1} U \\ &= (E_3 E_2 E_1)^{-1} U \end{aligned}$$

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

### Fast Method

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

$$\begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & -2 \\ -3 & 1 & 1 \end{bmatrix} \quad (1)$$

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

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

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

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

## Derivatives

$$\begin{aligned} \frac{df}{dx} f(g(x)) &= f'(g(x))g'(x) & \frac{d}{dx} ax^n &= nax^{n-1} & \frac{d}{dx} \ln_a(x) &= \frac{1}{x \ln(a)} \\ \frac{d}{dx} a^x &= \ln(a)a^x & \frac{d}{dx} \sin(x) &= \cos(x) & \frac{d}{dx} \cos(x) &= -\sin(x) \end{aligned}$$

## PA = LU Factorization

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

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 matrices
3. You'll have multiple  $P$  matrices, such that

$$PA = LU$$

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

## Solving

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

Example 1:  $A = LU$

$$\text{Let } \vec{b} = \begin{bmatrix} 2 \\ 1 \\ -1 \end{bmatrix}$$

1.  $L\vec{y} = \vec{b}$

$$\left[ \begin{array}{ccc|c} 1 & 0 & 0 & 2 \\ 2 & 1 & 0 & 1 \\ -3 & -\frac{7}{3} & 1 & -1 \end{array} \right] \rightarrow \vec{y} = \begin{pmatrix} 2 \\ -3 \\ -2 \end{pmatrix}$$

2.  $U\vec{x} = \vec{y}$

$$\left[ \begin{array}{ccc|c} 1 & 2 & -1 & 2 \\ 0 & -3 & 0 & -3 \\ 0 & 0 & -2 & -2 \end{array} \right] \rightarrow \vec{x} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

Example 2: -  $PA = LU$

$$\text{Let } \vec{b} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \text{ and } A = \begin{bmatrix} 2 & 1 & 5 \\ 4 & 4 & -4 \\ 1 & 3 & 1 \end{bmatrix}$$

1. Find  $P$  such that  $A$ 's pivots are the largest in magnitude per column.  
In this instance:

$$P = P_2 P_1 \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix}$$

2.  $L\vec{y} = P\vec{b}$

$$\left[ \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ \frac{1}{4} & 1 & 0 & 6 \\ \frac{1}{2} & -\frac{1}{2} & 1 & 5 \end{array} \right] \rightarrow \vec{y} = \begin{pmatrix} 0 \\ 6 \\ 8 \end{pmatrix}$$

3.  $U\vec{x} = \vec{y}$

$$\left[ \begin{array}{ccc|c} 4 & 4 & -4 & 0 \\ 0 & 2 & 2 & 6 \\ 0 & 0 & 8 & 8 \end{array} \right] \rightarrow \vec{x} = \begin{pmatrix} -1 \\ 2 \\ 1 \end{pmatrix}$$