



# Introduction to Numerical Analysis for Engineers

- Fundamentals of Digital Computing
  - Digital Computer Models
  - Convergence, accuracy and stability
  - Number representation
  - Arithmetic operations
  - Recursion algorithms
- Error Analysis
  - Error propagation – numerical stability
  - Error estimation
  - Error cancellation
  - Condition numbers

# Error Propagation

## Spherical Bessel Functions

### Differential Equation

$$x^2 \frac{d^2 y}{dx^2} + 2x \frac{dy}{dx} (x^2 - n(n+1)) y = 0$$

### Solutions

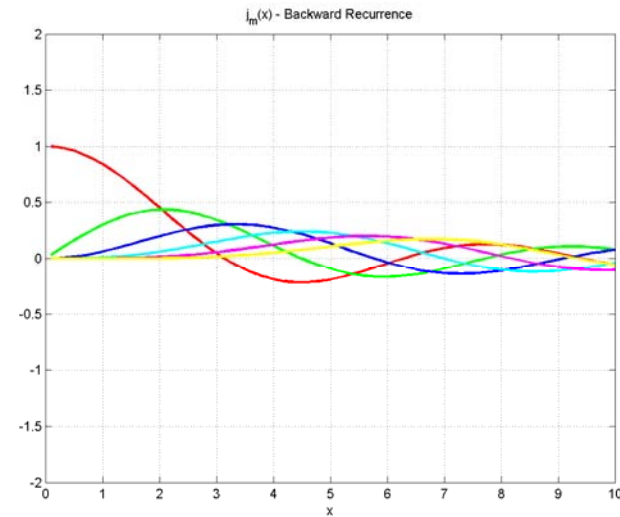
$$j_n(x) y_n(x)$$

$n$	$j_n(x)$	$y_n(x)$
0	$\frac{\sin x}{x}$	$-\frac{\cos x}{x}$
1	$\frac{\sin x}{x^2} - \frac{\cos x}{x}$	$-\frac{\cos x}{x^2} - \frac{\sin x}{x}$

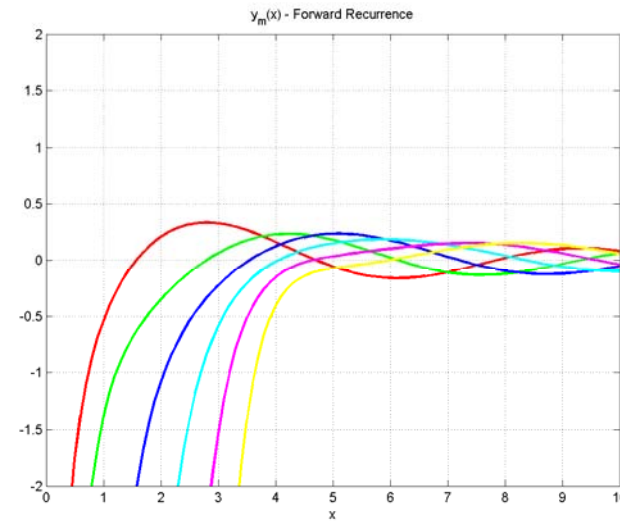
$$j_n(x) \rightarrow 0 \begin{cases} n \rightarrow \infty \\ x \rightarrow 0 \end{cases}$$

$$y_n(x) \rightarrow -\infty \begin{cases} n \rightarrow \infty \\ x \rightarrow 0 \end{cases}$$

$j_n(x)$



$y_n(x)$



# Error Propagation

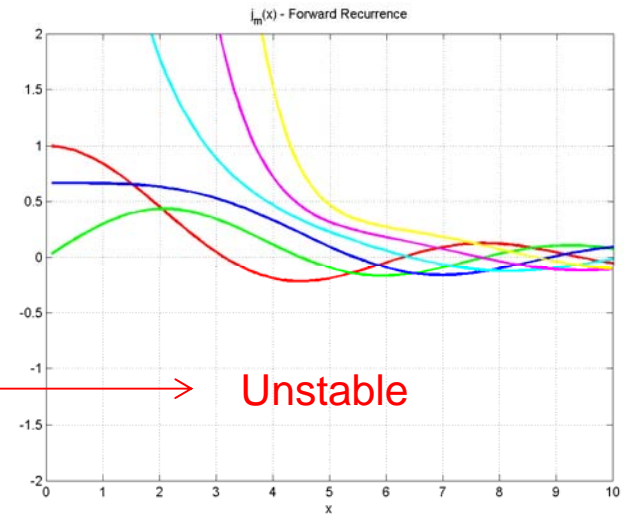
## Spherical Bessel Functions

### Forward Recurrence

$$j_{n+1}(x) = \frac{2n+1}{x} j_n(x) - j_{n-1}(x)$$

### Forward Recurrence

$$\frac{2n+1}{x} j_n(x) \simeq j_{n-1}(x) \leftarrow$$



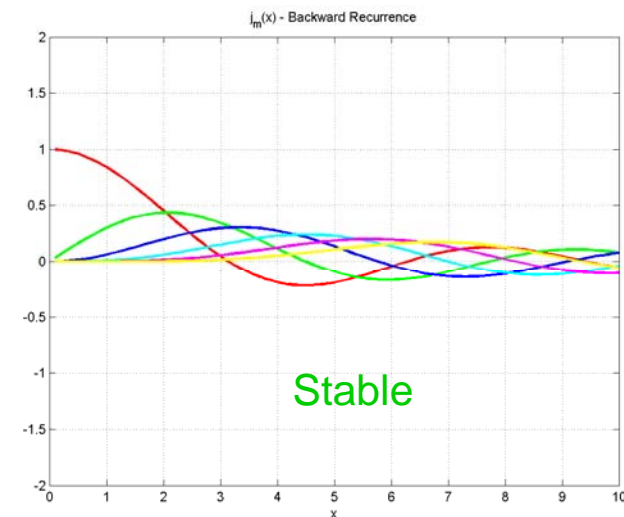
### Backward Recurrence

$$j_{n-1}(x) = \frac{2n+1}{x} j_n(x) - j_{n+1}(x)$$

### Miller's algorithm

$$j_N(x) = 1, \quad j_{N+1}(x) = 0, \quad j_0(x) = \frac{\sin x}{x}$$

$$N \sim x+20$$



# Error Propagation

## Euler's Method

### Differential Equation

$$\frac{dy}{dx} = f(x, y), \quad y_0 = p$$

### Example

$$f(x, y) = x \left( y = \frac{x^2}{2} + p \right)$$

### Discretization

$$x_n = nh$$

### Finite Difference (forward)

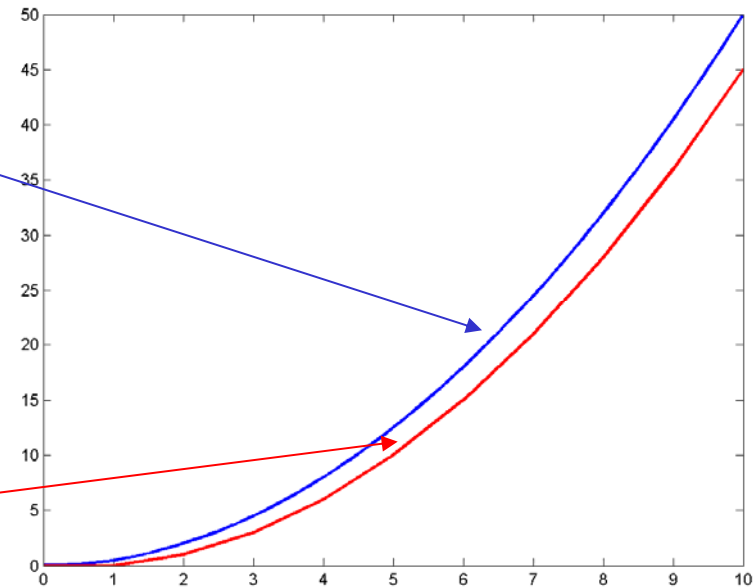
$$\frac{dy}{dx} \Big|_{x=x_n} \simeq \frac{y_{n+1} - y_n}{h}$$

### Recurrence

$$y_{n+1} = y_n + hf(nh, y)$$

### Central Finite Difference

$$\frac{dy}{dx} \Big|_{x=x_n} \simeq \frac{y_{n+1} - y_{n-1}}{2h}$$



euler.m



# Error Analysis

## Numerical Instability Example

### Evaluate Integral

$$y_n = \int_0^1 \frac{x^n}{x+5} dx, n = 0, 2 \dots \infty$$

### Recurrence Relation

$$y_n = \frac{1}{n} - 5y_{n-1}$$

### Proof

$$y_n + 5y_{n-1} = \int_0^1 \frac{x^n + 5x^{n-1}}{x+5} dx = \int_0^1 \frac{x^{n-1}(x+5)}{x+5} dx = \int_0^1 x^{n-1} dx = \frac{1}{n}$$

### 3-digit Recurrence

$$y_0 = \int_0^1 \frac{dx}{x+5} = [\log_e(x+5)]_0^1 = \log_e 6 - \log_e 5 = 0.182$$

$$y_1 = 1 - 5y_0 = 1 - 0.910 \simeq 0.0090$$

$$y_2 = 0.5 - 5y_1 \simeq 0.050$$

$$y_3 = 0.333 - 5y_2 \simeq 0.083 \quad > y_2 !!$$

$$y_4 = 0.25 - 5y_3 \simeq -0.165 \quad < 0 !!$$

### Backward Recurrence

$$y_{n-1} = \frac{1}{5n} - \frac{y_n}{5}$$

$$y_{10} \simeq y_9 \Rightarrow y_9 + 5y_9 = 0.1 \Rightarrow y_9 = 0.017$$

$$y_8 = 1/45 - y_9/5 = 0.019$$

$$y_7 = 1/40 - y_8/5 = 0.021$$

$$y_6 = 0.025$$

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$$y_1 = 0.088$$

$$y_0 = 0.182 \quad \text{Correct}$$

Exercise: Make MATLAB script

# Error Propagation

$$y = f(x_1, x_2, \dots, x_n)$$

Absolute Errors

$$\epsilon_1, \epsilon_2, \dots, \epsilon_n$$

$$\epsilon_y$$

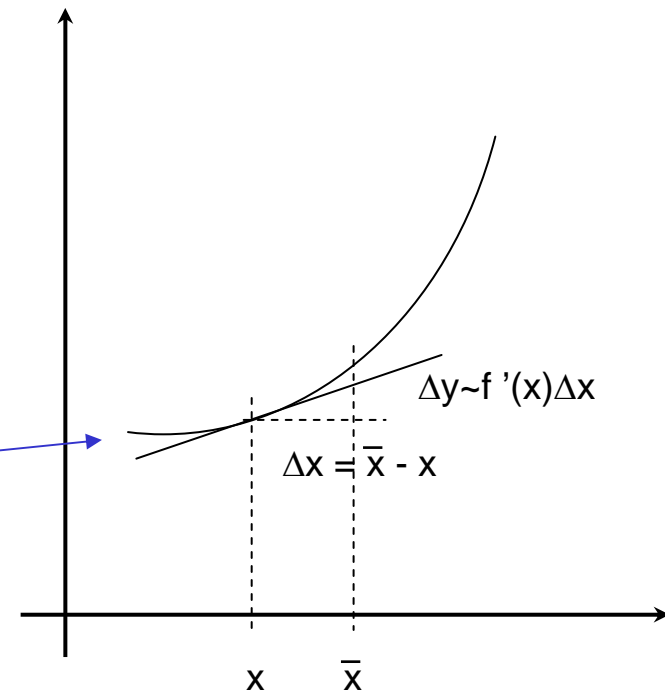
Function of one variable

$$y = f(x) \quad \bar{y} = f(\bar{x})$$

General Error Propagation Formula

$$\Delta y \simeq \sum_{i=1}^n \frac{\partial f(x_1, \dots, x_n)}{\partial x_i} \Delta x_i$$

$$\epsilon_y \leq \sum_{i=1}^n \left| \frac{\partial f(x_1, \dots, x_n)}{\partial x_i} \right| |\epsilon_i|$$





# Error Propagation Example

## Multiplication

$$y = x_1 x_2$$

$$\Rightarrow \log y = \log x_1 + \log x_2$$

$$\Rightarrow \frac{1}{y} \frac{\partial y}{\partial x_i} = \frac{1}{x_i}$$

$$\Rightarrow \frac{\partial y}{\partial x_i} = \frac{y}{x_i}$$

$\Rightarrow$

## Error Propagation Formula

$$\left| \frac{\Delta y}{y} \right| \leq \sum_{i=1}^2 \left| \frac{\Delta x_i}{x_i} \right|$$

$$\alpha_y \leq \sum_{i=1}^2 \alpha_i$$

$$y = x_1^{m_1} x_2^{m_2} \cdots x_n^{m_n}$$

$$\alpha_y \leq \sum_{i=1}^n |m_i| \alpha_i$$

Relative Errors Add for Multiplication



# Error Propagation Expectation of Errors

## Addition

$$y = x_1 + x_2 + \cdots + x_n$$

## Truncation

$$\Delta x_i = \bar{x}_i - x_i \leq 0$$

## Error Expectation

$$E(\Delta x_i) = -b^{-t}/2$$

$$E(\Delta y) = -nb^{-t}/2$$

## Rounding

$$E(\Delta x_i) = 0$$

$$E(\Delta y) = -nE(\Delta x_i) = 0$$

## Standard Error

$$E(\Delta_s y) \simeq \sqrt{\sum_{i=1}^n \left(\frac{\partial y}{\partial x_i}\right)^2 \epsilon_i^2}$$

$$y = x_1 + x_2 + \cdots + x_n$$

$$E(\Delta_s y) \simeq \sqrt{\sum_{i=1}^n \epsilon_i^2} = \sqrt{n}\epsilon$$

Standard Error better measure  
of expected errors





# Error Propagation Error Cancellation

## Function of one variable

$$y = f(x) = \sqrt{x^2 + 1} - x + 200; \quad x = 100 \pm 4$$

$$z_1 = \sqrt{x^2 + 1} \quad \epsilon_1 = 4$$

$$z_2 = 200 - x \quad \epsilon_2 = 4$$

$$y = z_1 + z_2$$

Max. error  $E(\Delta y) = 8$

Stand. error  $E(\Delta_s y) = 4\sqrt{2} = 5.6$

$$\frac{\partial z_1}{\partial x} = \frac{x}{\sqrt{x^2 + 1}}, \quad \frac{\partial z_2}{\partial x} = -1$$

$$\Delta y \simeq \frac{dz_1}{dx} \Delta x + \frac{dz_2}{dx} \Delta x$$

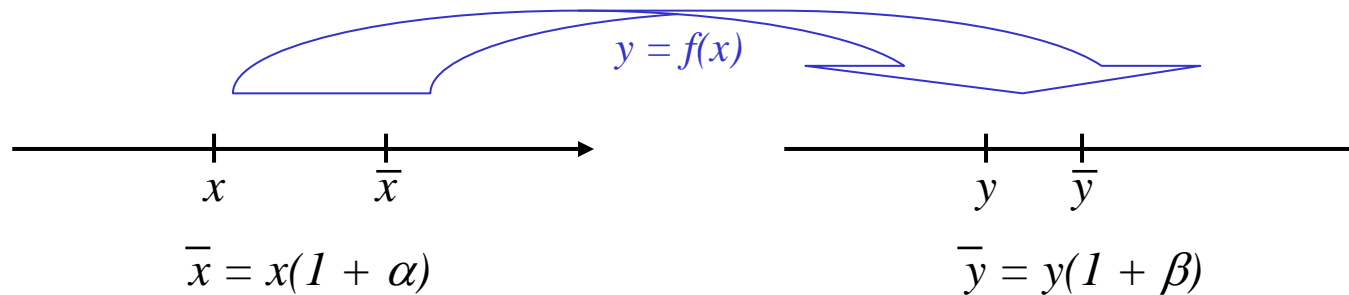
$$= \left( \frac{x}{\sqrt{x^2 + 1}} - 1 \right) \Delta x \simeq_{x \gg 1} \frac{-1}{x^2} \Delta x$$

**Error cancellation**

$$x = 100 \pm 4 \Rightarrow |\Delta y| \leq 4 \cdot 10^{-4} \leq 0.5 \cdot 10^{-3}$$

$$y = 200.005 \pm 0.5 \cdot 10^{-3}$$

# Error Propagation Condition Number



## Problem Condition Number

$$\begin{aligned}
 K_P &= \frac{|\beta|}{|\alpha|} \\
 &= \left| \frac{f(\bar{x}) - f(x)}{f(x)} \right| / \left| \frac{\bar{x} - x}{x} \right| \\
 &= \left| \frac{f(\bar{x}) - f(x)}{\bar{x} - x} \right| \times \left| \frac{x}{f(x)} \right| \\
 &\approx \left| x \frac{f'(x)}{f(x)} \right|
 \end{aligned}$$

$$K_P \gg 1$$

Problem ill-conditioned

## Error cancellation example

$$y = f(x) = \sqrt{x^2 + 1} - x + 200; \quad x = 100 \pm 4$$

$$K_P = \left| 100 \frac{-10^{-4}}{200.005} \right| = 0.5 \cdot 10^{-4}$$

Well-conditioned problem



# Error Propagation Condition Number

## Problem Condition Number

$$y = \sqrt{x^2 + 1} - x$$

$$x = 100 \Rightarrow y = 0.5 \cdot 10^{-2}$$

$$y' = \frac{x}{\sqrt{x^2 + 1}} - 1 \simeq -\frac{1}{x^2} = -10^{-4}$$

$$K_P = \left| 100 \frac{-1 \cdot 10^{-4}}{0.5 \cdot 10^{-4}} \right| = 2.0$$

4 Significant Digits

$$\begin{aligned} \bar{y} &= \sqrt{0.1 \cdot 10^5 + 1} - 0.1 \cdot 10^3 \\ &= \sqrt{(0.1000 + 0.00001) \cdot 10^5} - 0.1 \cdot 10^3 = 0 \end{aligned}$$

$$|\beta| = \left| \frac{\bar{y} - y}{y} \right| = \frac{0.5 \cdot 10^{-2}}{0.5 \cdot 10^{-2}} = 1$$

$$|\alpha| = \left| \frac{\bar{x} - x}{x} \right| \leq \frac{1}{2} 10^{1-t} \simeq \frac{1}{2} 10^{-3}$$

Algorithm Condition Number

$$K_A = \frac{|\beta|}{|\alpha|} \simeq 2000$$

$K_A$  is algorithm condition number, which may be much larger than the  $K_P$  due to limited number representation.

Solution

- Higher precision
- Rewrite algorithm

Well-conditioned Algorithm

$$y = \frac{1}{\sqrt{x^2 + 1} + x}$$

$$\bar{y} = \frac{1}{0.1 \cdot 10^3 + 0.1 \cdot 10^3} = 0.5 \cdot 10^{-2}$$

$$|\beta| \simeq 0 \Rightarrow K_A \simeq 0 \ll 1$$