

# Numerical Methods and Optimization

Topic 1:

Introduction to Numerical Methods

Lectures 1-4:

# Lecture 1

## Introduction to Numerical Methods

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- What are **NUMERICAL METHODS**?
- Why do we need them?
- Topics covered in **NMO**.

**Reading Assignment:** Pages 3-10 of textbook

# Numerical Methods

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## **Numerical Methods:**

Algorithms that are used to obtain numerical solutions of a mathematical problem.

## **Why do we need them?**

1. No analytical solution exists,
2. An analytical solution is difficult to obtain or not practical.

# What do we need?

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## **Basic Needs in the Numerical Methods:**

- **Practical:**

- Can be computed in a reasonable amount of time.

- **Accurate:**

- Good approximate to the true value,
  - Information about the approximation error (Bounds, error order,... ).

# Outlines of the Course

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- ❑ Number Representation
- ❑ Approximate solution of nonlinear Equations
- ❑ Solution of linear Equations (Direct methods)
- ❑ Solution of linear Equations (Iterative methods)
- ❑ Polynomial Interpolation
- ❑ Least Squares approximation
- ❑ Numerical Integration

# Solution of Nonlinear Equations

- Some simple equations can be solved analytically:

$$x^2 + 4x + 3 = 0$$

$$\text{Analytic solution } roots = \frac{-4 \pm \sqrt{4^2 - 4(1)(3)}}{2(1)}$$

$$x = -1 \text{ and } x = -3$$

- Many other equations have no analytical solution:

$$\left. \begin{array}{l} x^9 - 2x^2 + 5 = 0 \\ x = e^{-x} \end{array} \right\} \text{No analytic solution}$$

# Methods for Solving Nonlinear Equations

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- o **Bisection Method**
- o **Newton-Raphson Method**
- o **Secant Method**

# Solution of Systems of Linear Equations

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$$x_1 + x_2 = 3$$

$$x_1 + 2x_2 = 5$$

We can solve it as :

$$x_1 = 3 - x_2, \quad 3 - x_2 + 2x_2 = 5$$

$$\Rightarrow x_2 = 2, \quad x_1 = 3 - 2 = 1$$

What to do if we have

1000 equations in 1000 unknowns.



# Cramer's Rule is Not Practical

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Cramer's Rule can be used to solve the system :

$$x_1 = \frac{\begin{vmatrix} 3 & 1 \\ 5 & 2 \\ 1 & 1 \\ 1 & 2 \end{vmatrix}}{\begin{vmatrix} 1 & 3 \\ 1 & 5 \\ 1 & 1 \\ 1 & 2 \end{vmatrix}} = 1, \quad x_2 = \frac{\begin{vmatrix} 1 & 3 \\ 1 & 5 \\ 1 & 1 \\ 1 & 2 \end{vmatrix}}{\begin{vmatrix} 1 & 3 \\ 1 & 5 \\ 1 & 1 \\ 1 & 2 \end{vmatrix}} = 2$$

But Cramer's Rule is not practical for large problems.

To solve  $N$  equations with  $N$  unknowns, we need  $(N+1)(N-1)N!$  multiplications.

To solve a 30 by 30 system,  $2.3 \times 10^{35}$  multiplications are needed.

A super computer needs more than  $10^{20}$  years to compute this.

# Methods for Solving Systems of Linear Equations

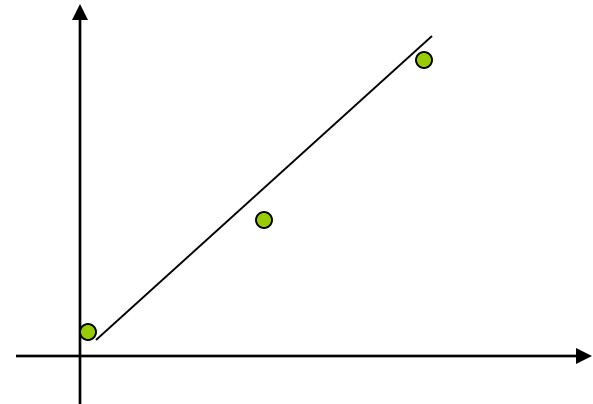
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- o **Naive Gaussian Elimination**
- o **Gaussian Elimination with Scaled Partial Pivoting**
- o **Algorithm for Tri-diagonal Equations**

# Curve Fitting

- Given a set of data:

x	0	1	2
y	0.5	10.3	21.3

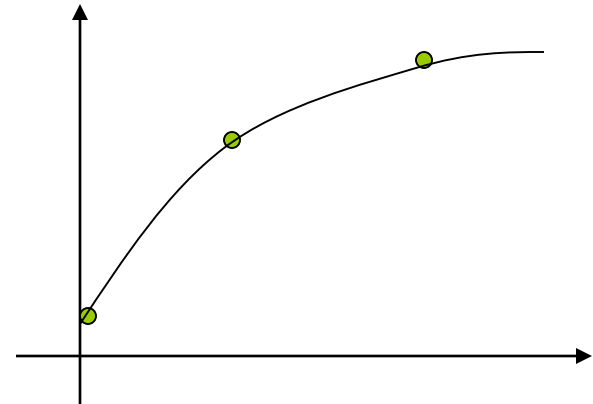


- Select a curve that best fits the data. One choice is to find the curve so that the sum of the square of the error is minimized.

# Polynomial Interpolation

□ Given a set of data:

$x_i$	0	1	2
$y_i$	0.5	10.3	15.3



□ Find a polynomial  $P(x)$  whose graph passes through all tabulated points.

$$y_i = P(x_i) \quad \text{if } x_i \text{ is in the table}$$

# Methods for Curve Fitting

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- o **Least Squares**
  - o **Linear Regression**
  - o **Nonlinear Least Squares Problems**
- o **Interpolation**
  - o **Newton Polynomial Interpolation**
  - o **Lagrange Interpolation**

# Integration

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- Some functions can be integrated analytically:

$$\int_1^3 x dx = \frac{1}{2} x^2 \Big|_1^3 = \frac{9}{2} - \frac{1}{2} = 4$$

But many functions have no analytical solutions :

$$\int_0^a e^{-x^2} dx = ?$$

# Methods for Numerical Integration

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- o **Upper and Lower Sums**
- o **Trapezoid Method**
- o **Romberg Method**
- o **Gauss Quadrature**

# Summary

- ▣ **Numerical Methods:**  
Algorithms that are used to obtain numerical solution of a mathematical problem.
- ▣ **We need them when**  
No analytical solution exists or it is difficult to obtain it.

## Topics Covered in the Course

- ▣ Solution of Nonlinear Equations
- ▣ Solution of Linear Equations
- ▣ Curve Fitting
  - Least Squares
  - Interpolation
- ▣ Numerical Integration



## Lecture 2

# Number Representation and Accuracy



- ❑ Number Representation
- ❑ Normalized Floating Point Representation
- ❑ Significant Digits
- ❑ Accuracy and Precision
- ❑ Rounding and Chopping

**Reading Assignment:** Chapter 3

# Representing Real Numbers

- You are familiar with the decimal system:

$$312.45 = 3 \times 10^2 + 1 \times 10^1 + 2 \times 10^0 + 4 \times 10^{-1} + 5 \times 10^{-2}$$

- Decimal System: Base = 10 , Digits (0,1,...,9)

- Standard Representations:

±	3	1	2	.	4	5
sign	integral				fraction	
	part				part	

# Normalized Floating Point Representation

## □ Normalized Floating Point Representation:

$$\begin{array}{ccccc} \pm & d. & f_1 & f_2 & f_3 & f_4 & \times 10^{\pm n} \\ \text{sign} & & \text{mantissa} & & & & \text{exponent} \end{array}$$

$d \neq 0$ ,  $\pm n$  : signed exponent

- Scientific Notation: Exactly one non-zero digit appears before decimal point.
- Advantage: Efficient in representing very small or very large numbers.

# Binary System

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▣ Binary System: Base = 2, Digits {0,1}

$$\begin{array}{ccccc} \pm & 1. & f_1 & f_2 & f_3 & f_4 & \times & 2^{\pm n} \\ \text{sign} & & \text{mantissa} & & & & & \uparrow \\ & & & & & & & \text{signed exponent} \end{array}$$

$$(1.101)_2 = (1 + 1 \times 2^{-1} + 0 \times 2^{-2} + 1 \times 2^{-3})_{10} = (1.625)_{10}$$

# Fact

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- Numbers that have a finite expansion in one numbering system may have an infinite expansion in another numbering system:

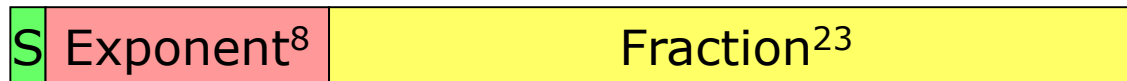
$$(1.1)_{10} = (1.000110011001100\dots)_2$$

- You can never represent 1.1 exactly in binary system.

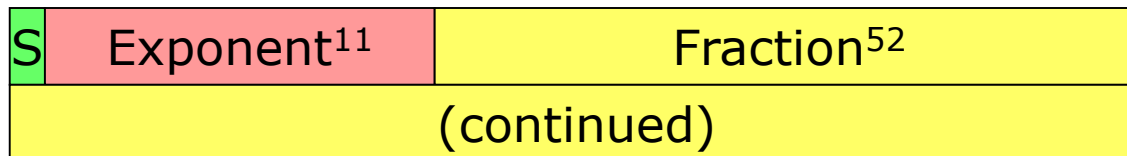
# IEEE 754 Floating-Point Standard

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- ❑ Single Precision (32-bit representation)
  - 1-bit Sign + 8-bit Exponent + 23-bit Fraction



- ❑ Double Precision (64-bit representation)
  - 1-bit Sign + 11-bit Exponent + 52-bit Fraction



# Significant Digits

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- Significant digits are those digits that can be used with confidence.

- Single-Precision: 7 Significant Digits

$$1.175494... \times 10^{-38} \text{ to } 3.402823... \times 10^{38}$$

- Double-Precision: 15 Significant Digits

$$2.2250738... \times 10^{-308} \text{ to } 1.7976931... \times 10^{308}$$

# Remarks

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- Numbers that can be exactly represented are called machine numbers.
- Difference between machine numbers is not uniform
- Sum of machine numbers is not necessarily a machine number



# Calculator Example

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- Suppose you want to compute:

$$3.578 * 2.139$$

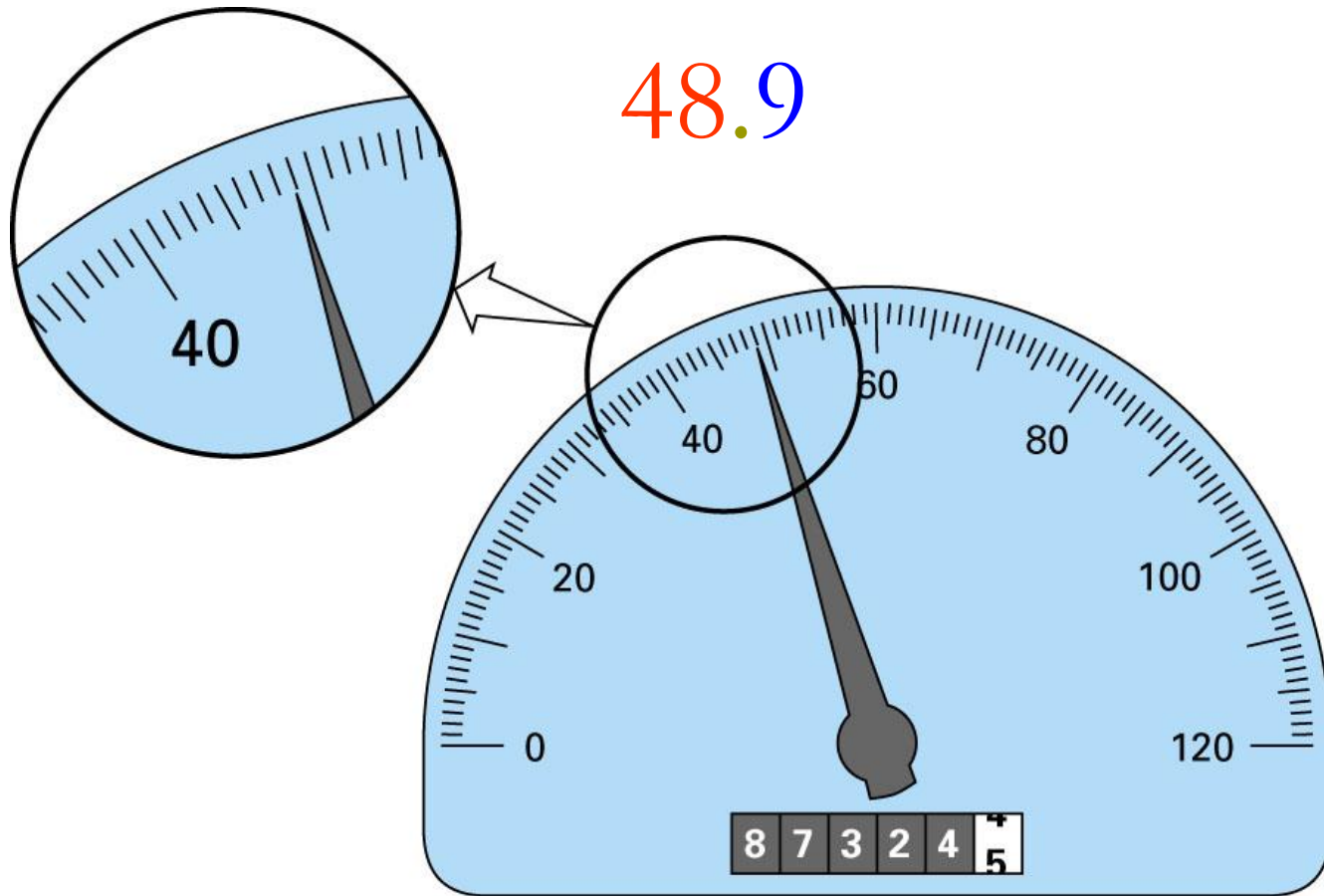
using a calculator with two-digit fractions

$$\boxed{3.57} * \boxed{2.13} = \boxed{7.60}$$

**True answer:**

**7.653342**

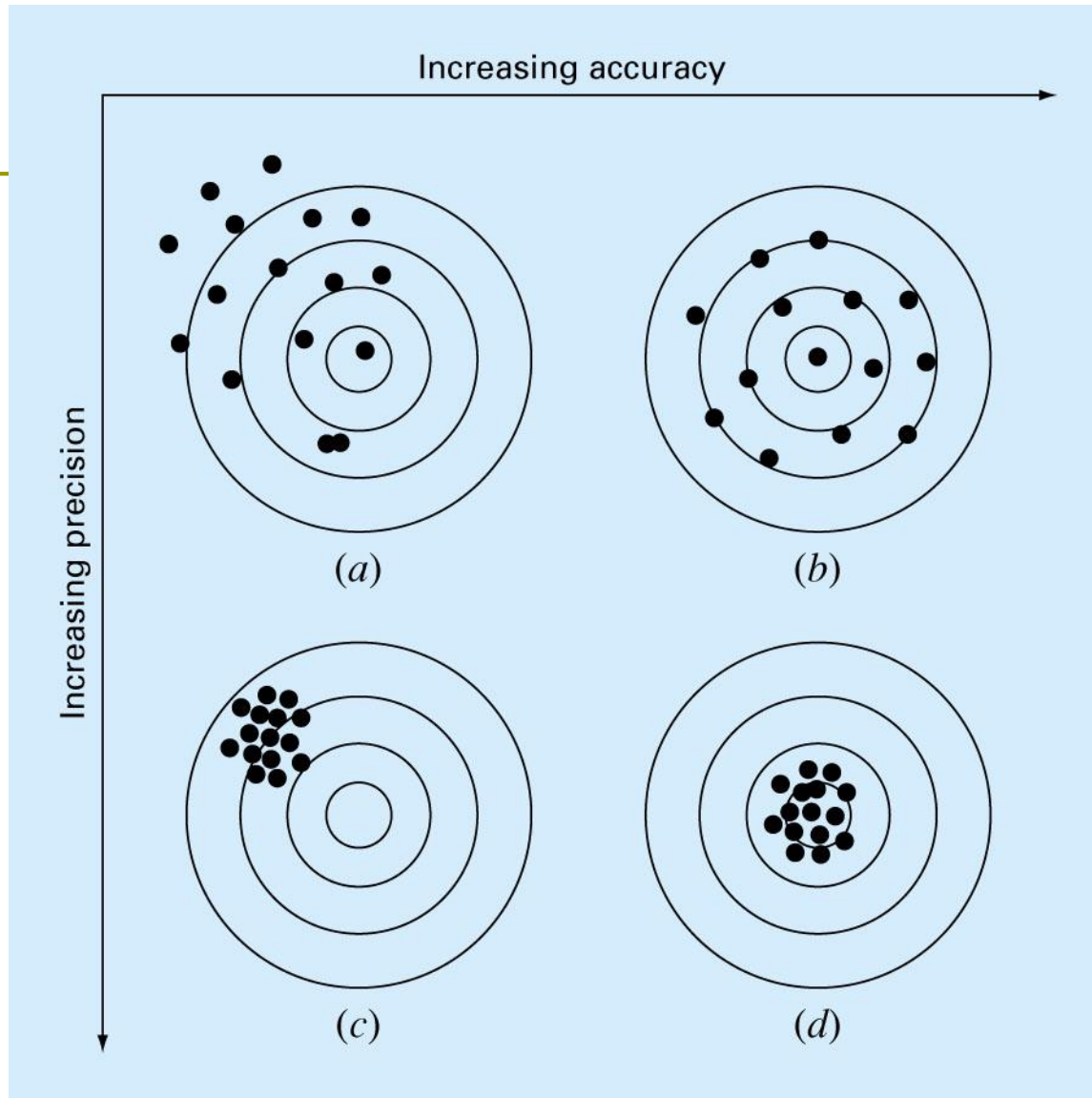
# Significant Digits - Example



# Accuracy and Precision

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- ❑ Accuracy is related to the closeness to the true value.
- ❑ Precision is related to the closeness to other estimated values.



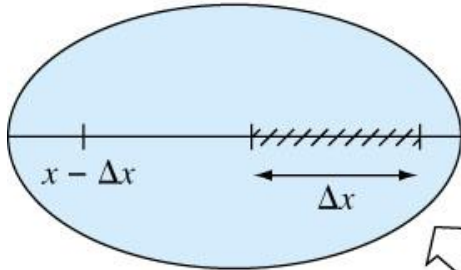
# Rounding and Chopping

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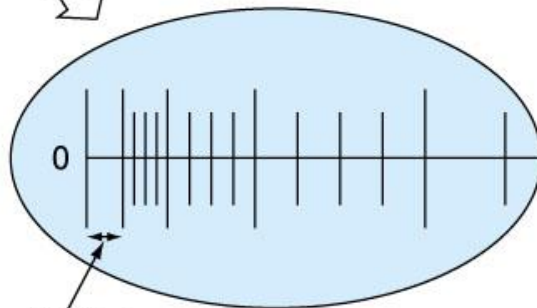
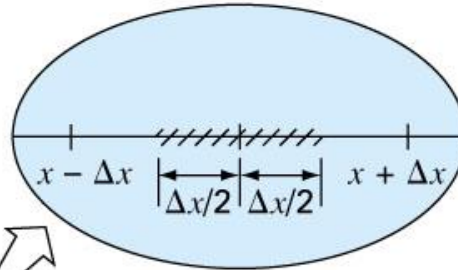
- ❑ Rounding: Replace the number by the nearest machine number.
- ❑ Chopping: Throw all extra digits.

# Rounding and Chopping

Chopping



Rounding



Underflow "hole"  
at zero

# Error Definitions — True Error

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Can be computed if the true value is known:

Absolute True Error

$$E_t = | \text{true value} - \text{approximation} |$$

Absolute Percent Relative Error

$$\varepsilon_t = \left| \frac{\text{true value} - \text{approximation}}{\text{true value}} \right| * 100$$

# Error Definitions — Estimated Error

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When the true value is not known:

Estimated Absolute Error

$$E_a = | \text{current estimate} - \text{previous estimate} |$$

Estimated Absolute Percent Relative Error

$$\mathcal{E}_a = \left| \frac{\text{current estimate} - \text{previous estimate}}{\text{current estimate}} \right| * 100$$



# Notation

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We say that the estimate is correct to  $n$  decimal digits if:

$$|\text{Error}| \leq 10^{-n}$$

We say that the estimate is correct to  $n$  decimal digits **rounded** if:

$$|\text{Error}| \leq \frac{1}{2} \times 10^{-n}$$

# Summary

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## □ Number Representation

Numbers that have a finite expansion in one numbering system may have an infinite expansion in another numbering system.

## □ Normalized Floating Point Representation

- Efficient in representing very small or very large numbers,
- Difference between machine numbers is not uniform,
- Representation error depends on the number of bits used in the mantissa.