# 10.34: Numerical Methods Applied to Chemical Engineering

Lecture I: Organization, Numerical Error, Basics of Linear Algebra

- Purposes of the course:
  - Ensure that you are aware of the wide range of easily accessible numerical methods that will be useful in your thesis research, at practice school, and in your career.
  - Make you confident in your ability to look up and apply additional methods when you need them.
  - Help you become familiar with MATLAB, other convenient numerical software, and with simple programming/debugging techniques.
  - Give you an understanding of how common numerical algorithms work and why they sometimes produce unexpected results.

- Resources:
  - Course website details on grading, homework policy, and homework submission guidelines.
  - Textbook Beers, "Numerical Methods for Chemical Engineering". Notes will be placed on Course website. Additional text references are given in the syllabus.
  - MATLAB tutorials
  - Peers you are encouraged to discuss the course material, programming, and the homework with your colleagues. Be aware of the homework policy outlined in the syllabus, however.
  - TAs and instructors we are here to help you, and available for meetings, usually within 24 hours.

- When to stop:
  - The homework for the course should require 9 hours per week on average – perhaps a little more early on if you are not proficient with MATLAB.
  - Sometimes you may find a homework problem is consuming an inordinate amount of time even after you have asked for help.
  - If this happens, just turn in what you have completed with a note indicating that you know your solution is incomplete, details about what you think went wrong, and what you think a correct solution would look like.

- Linear algebra
- Solutions of nonlinear equations
- Optimization
- Initial value problems
- Differential-algebraic equations
- Boundary value problems
- Partial differential equations
- Probability theory
- Monte Carlo methods
- Stochastic chemical kinetics

## Numerical Methods

- Motivation:
  - Most real engineering problems do not have an exact solution. Even if there is an exact solution. Can it be evaluated exactly?
  - Application of computational problem solving methodologies can lead to transformative (as opposed to incremental) engineering solutions.
- Algorithms to solve problems numerically should be:
  - clear
  - concise
  - able to solve the problem robustly
  - use realistic amount of resources
  - execute in a realistic amount of time

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- Virtually all computer problem solving is done approximately. It is essential to quantify the error in these calculations.
  - Example: representation of numbers

 $\pi = 3.141592653589\ldots$ 

1100100100001111110110101

significand (24 bits) exponent (8 bits)  $1 + \sum_{n=1}^{p-1} \operatorname{bit}_n \times 2^{-n} \times 2^e$ 

• Example: calculating the square root  $\sqrt{s}$ 

 $x^2 - s = 0$ 

Babylonian method (iterative solution):

$$x_{n+1} = \frac{1}{2} \left( x_n + \frac{s}{x_n} \right)$$

- Overflow/underflow exceeding the largest/smallest representable number
  - Example:  $I.3 \times 10^{45} (nm)^3 = I.3 \times 10^9 (km)^3$
  - Solution: rescaling
- Truncation:
  - Computers have a finite amount of memory/time to work with. Most algorithms work within these constraints to return answers which are accurate to within some tolerance.
  - Solution: the design of algorithms that quickly minimize truncation error
  - Example: Leibniz vs. Newton

$$\sum_{n=0}^{\infty} \frac{(-1)^n}{2n+1} = \frac{\pi}{4}$$

$$\frac{1}{2}\sum_{n=0}^{\infty} \frac{2^n n!^2}{(2n+1)!} = \frac{\pi}{4}$$

- Truncation (cont.):
  - Example: Leibniz vs. Newton

$$\sum_{n=0}^{N} \frac{(-1)^n}{2n+1} = \frac{\pi}{4} \qquad \qquad \frac{1}{2} \sum_{n=0}^{N} \frac{2^n n!^2}{(2n+1)!} = \frac{\pi}{4}$$

N	Leibniz	Newton			
I	0.66667	0.66667			
2	0.86667	0.73333			
3	0.72381	0.76190			
5	0.74401	0.78038			
10	0.80808	0.78528			

0.78540...

• Absolute error:

$$\epsilon_{\text{abs.}} = |x_{\text{exact}} - x_{\text{approx.}}|$$
  
Relative error:  
 $\epsilon_{\text{rel.}} = \frac{|x_{\text{exact}} - x_{\text{approx.}}|}{|x_{\text{exact}}|}$ 

- Truncation (cont.):
  - Example:  $2 \times 10^{-4} + 1 \times 10^{-13} = ?$  with 8 digit accuracy
    - Estimate the absolute error in this calculation.
    - Estimate the relative error in this calculation.
- Quantifying and minimizing numerical error is a key aspect developing numerical algorithms.
- Even simple calculations introduce numerical errors.
  - Those errors can compound and magnify. We will see how shortly.

## Linear Algebra

- Primarily concerned with the solutions of systems of linear equations
  - Is there a solution?
  - If there is a solution, is it a unique?
  - Is it possible to find the solution or family of solutions?
- Chemical engineering example: mass balances

### Linear Algebra

- Row-view:
  - Each row in the system of equations describes a line.
  - The solution represents the intersection of these lines.
  - For dimensions higher than 2, the solution is an intersection of other linear manifolds
  - How many solutions does the equation: ax=b, have?

$$\dot{m}_1 + \dot{m}_2 = 3$$
  
 $-2\dot{m}_1 + \dot{m}_2 = 0$ 





- Column view:
  - Each column in the system of equations describes a vector.
  - The solution represents the correct weighting of these vectors.
  - While conceptually more difficult, the column view is easier to extend to arbitrarily high dimensions. You will see why later.

$$\dot{m}_1 \left( \begin{array}{c} 1 \\ -2 \end{array} \right) + \dot{m}_2 \left( \begin{array}{c} 1 \\ 1 \end{array} \right) = \left( \begin{array}{c} 3 \\ 0 \end{array} \right)$$







### Solving Systems of Equations

$$ax = b \Rightarrow x = a^{-1}b$$

$$Ax = b \Rightarrow x = A^{-1}b$$

#### In MATLAB:

$$x = A \setminus b$$

- Scalars:
  - Just single numbers!
  - Set of all real numbers,  ${\mathbb R}$
  - Set of all complex numbers,  $\mathbb C$

• 
$$i = \sqrt{-1}$$

- If  $z\in \mathbb{C}$  , then  $z=a+ib\,$  with  $a,b\in \mathbb{R}$
- Complex conjugate:  $\bar{z} = a ib$
- Magnitude:  $|z| = \sqrt{z\bar{z}}$
- $\mathbb{R} \subset \mathbb{C}$

- Vectors:
  - Ordered sets of numbers:  $(x_1, x_2, \ldots x_N)$
  - Set of all real vectors with dimension N,  $\mathbb{R}^N$
  - Addition:

$$\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_N \end{pmatrix} + \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_N \end{pmatrix} = \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \\ \vdots \\ x_N + y_N \end{pmatrix}$$

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• Multiplication by scalar:  

$$c(x_1 \ x_2 \ \dots \ x_N) = (cx_1 \ cx_2 \ \dots \ cx_N)$$
• Transpose:  

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_N \end{pmatrix} \quad \mathbf{x}^T = (x_1 \ x_2 \ \dots \ x_N)$$

N

- Vectors:
  - Scalar product:  $\mathbf{x} \cdot \mathbf{y} = \sum x_i y_i$

• Norm: 
$$\|\mathbf{x}\|_p = \left(\sum_{i=1}^N |x_i|^p\right)^{1/p}$$

- Properties:
  - Non-negative:  $\|\mathbf{x}\|_p \ge 0$
  - If  $\|\mathbf{x}\|_p = 0$  , then  $\mathbf{x} = 0$
  - $||c\mathbf{x}||_p = |c|||\mathbf{x}||_p$
  - $|\mathbf{x} \cdot \mathbf{y}| \le ||\mathbf{x}||_p ||\mathbf{y}||_q$  with  $p, q > 0, \ 1/p + 1/q = 1$
  - $\|\mathbf{x} + \mathbf{y}\|_p \le \|\mathbf{x}\|_p + \|\mathbf{y}\|_p$

- Vectors:
  - • norm:  $\|\mathbf{x}\|_{\infty} = \max_{i} |x_{i}|$
  - Examples of norms:

 $\mathbf{x} = (\sqrt{2}/2, \sqrt{2}/2)$ 

- $\|\mathbf{x}\|_1 =$
- $\|\mathbf{x}\|_2 =$
- $\bullet \|\mathbf{x}\|_{\infty} =$
- $\|\mathbf{x}\|_{\infty} \leq \|\mathbf{x}\|_{2} \leq \|\mathbf{x}\|_{1}$
- Families of vectors with the same norm: I-norm, 2-norm, ∞-norm

$$\|\mathbf{x}\|_p = \left(\sum_{i=1}^N |x_i|^p\right)^{1/p}$$

- Vectors:
  - $\infty$ -norm:  $\|\mathbf{x}\|_{\infty} = \max_{i} |x_i|$
  - Examples of norms:

 $\mathbf{x} = (\sqrt{2}/2, \sqrt{2}/2)$ 

- $\|\mathbf{x}\|_1 =$
- $\|\mathbf{x}\|_2 =$
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- $\|\mathbf{x}\|_{\infty} \leq \|\mathbf{x}\|_{2} \leq \|\mathbf{x}\|_{1}$
- Families of vectors with the same norm: I-norm, 2-norm, ∞-norm

$$\|\mathbf{x}\|_{p} = \left(\sum_{i=1}^{N} |x_{i}|^{p}\right)^{1/p}$$

• Vectors:

• Comparing vectors with norm metrics:

• 
$$\|\mathbf{x} - \mathbf{y}\|_2 \ge 0$$

• If 
$$\|\mathbf{x} - \mathbf{y}\|_2 = 0$$
 , then  $\mathbf{x} = \mathbf{y}$ 

• 
$$\|\mathbf{x} - \mathbf{y}\|_2 \le \|\mathbf{x} - \mathbf{v}\|_2 + \|\mathbf{y} - \mathbf{v}\|_2$$

- Calculating norms in MATLAB:
  - norm( x, p ), norm( x, Inf )
- How many operations to compute the norm?
- How can I measure relative and absolute error for vectors?



• Vectors:

• Comparing vectors with norm metrics:

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 , then  $\mathbf{x} = \mathbf{y}$ 

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$$\|\mathbf{x} - \mathbf{y}\|_2 \le \|\mathbf{x} - \mathbf{v}\|_2 + \|\mathbf{y} - \mathbf{v}\|_2$$

- Calculating norms in MATLAB:
  - norm( x, p ), norm( x, Inf )
- How many operations to compute the norm?
- The relative and absolute error in a vector:



- Vectors:
  - What mathematical object is the equivalent of an infinite dimensional vector?

- Vectors:
  - What mathematical object is the equivalent of an infinite dimensional vector?

- Matrices:
  - Ordered sets of numbers:  $\mathbf{A} = \begin{pmatrix} A_{11} & A_{12} & \dots & A_{1M} \\ A_{21} & A_{22} & \dots & A_{2M} \\ \vdots & \vdots & \ddots & \vdots \\ A_{N1} & A_{N2} & \dots & A_{NM} \end{pmatrix}$
  - Set of all real matrices with N rows and M columns,  $\mathbb{R}^{N imes M}$
  - Addition:  $\mathbf{C} = \mathbf{A} + \mathbf{B} \Rightarrow C_{ij} = A_{ij} + B_{ij}$
  - Multiplication by scalar:  $\mathbf{C} = c\mathbf{A} \Rightarrow C_{ij} = cA_{ij}$
  - Transpose:  $\mathbf{C} = \mathbf{A}^T \Rightarrow C_{ij} = A_{ji}$
  - Trace (square matrices):  $Tr \mathbf{A} = \sum_{i=1}^{N} A_{ii}$

- Matrices:
  - Matrix-vector product:  $\mathbf{y} = \mathbf{A}\mathbf{x} \Rightarrow y_i = \sum A_{ij}x_j$
  - Matrix-matrix product:  $\mathbf{C} = \mathbf{AB} \Rightarrow C_{ij} = \sum A_{ik} B_{kj}$

 $\mathcal{M}$ 

i=1

M

k=1

- Properties:
  - no commutation in general:  $\mathbf{AB} \neq \mathbf{BA}$
  - association:  $\mathbf{A}(\mathbf{BC}) = (\mathbf{AB})\mathbf{C}$
  - distribution: A(B + C) = AB + AC
  - transposition:  $(\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T$
  - inversion:  $\mathbf{A}^{-1}\mathbf{A} = \mathbf{A}\mathbf{A}^{-1} = \mathbf{I}$  if  $\det(\mathbf{A}) \neq 0$

- Matrices:
  - Matrix-matrix product:
    - Vectors are matrices too:

• 
$$\mathbf{x} \in \mathbb{R}^N$$
  $\mathbf{x} \in \mathbb{R}^{N \times 1}$ 

• 
$$\mathbf{y}^T \in \mathbb{R}^N \ \mathbf{y}^T \in \mathbb{R}^{1 \times N}$$

• What is: 
$$\mathbf{y}^T \mathbf{x}$$
 ?

$$\mathbf{C} = \mathbf{AB} \Rightarrow C_{ij} = \sum_{k=1}^{M} A_{ik} B_{kj}$$

- Matrices:
  - Matrix-matrix product:
    - Vectors are matrices too:

• 
$$\mathbf{x} \in \mathbb{R}^N$$
  $\mathbf{x} \in \mathbb{R}^{N \times 1}$ 

• 
$$\mathbf{y}^T \in \mathbb{R}^N \ \mathbf{y}^T \in \mathbb{R}^{1 \times N}$$

• What is: 
$$\mathbf{y}^T \mathbf{x}$$
 ?

- Matrices:
  - Dyadic product:  $\mathbf{A} = \mathbf{x}\mathbf{y}^T = \mathbf{x} \otimes \mathbf{y} \Rightarrow A_{ij} = x_i y_j$ 
    - Determinant (square matrices only):  $\det(\mathbf{A}) = \sum_{j=1}^{N} (-1)^{i+j} A_{ij} M_{ij}(\mathbf{A})$  $M_{ij}(\mathbf{A}) =$

$$\det \begin{pmatrix} A_{11} & A_{12} & \dots & A_{1(j-1)} & A_{1(j+1)} & \dots & A_{1N} \\ A_{21} & A_{22} & \dots & A_{2(j-1)} & A_{2(j+1)} & \dots & A_{2N} \\ \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ A_{(i-1)1} & A_{(j-1)2} & \dots & A_{(i-1)(j-1)} & A_{(i-1)(j+1)} & \dots & A_{(i-1)N} \\ A_{(i+1)1} & A_{(j+1)2} & \dots & A_{(i+1)(j-1)} & A_{(i+1)(j+1)} & \dots & A_{(i+1)N} \\ \vdots & \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ A_{N1} & A_{N2} & \dots & A_{N(j-1)} & A_{N(j+1)} & \dots & A_{NN} \end{pmatrix}$$
  
• 
$$\det(c) = c$$

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