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			
Ι	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:

$$\epsilon_{abs.} = |x_{exact} - x_{approx.}|$$

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

- 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 $z=a+ib\,\, {\rm with}\,\, a,b\in \mathbb{R}$
- 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:
 - ∞ -norm: $\|\mathbf{x}\|_{\infty} = \max_{i} |x_i|$
 - Examples of norms:

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

- $\|\mathbf{x}\|_1 =$
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- 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?

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 - 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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