

# Agenda

9	10/17	Midterm review	10/19	<b>Midterm</b>	
10	10/24	Gaussian elimination (2.1)	10/26	Pivoting (2.2)	<b>HW4</b>
11	10/31	Pivoting (2.2)	11/2	Structured system (2.3)	
12	11/7	Factorization (8.1)	11/9		
13	11/14	Iterative (8.4)	11/16		<b>HW5</b>
14	11/21	<b>Fall break</b>	11/23	<b>Fall break</b>	
15	11/28	Eigenvalue (8.2)	11/30	Power method (8.3)	
16	12/5	Topic TBD	12/8	Review	<b>HW6</b>
	12/12	<b>2:00-4:45pm Final Exam</b>			

# Overview

Solving an algebraic linear system

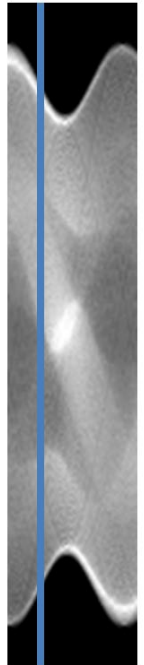
$$Ax = b$$

for the unknown vector  $x$ , when the **coefficient matrix**  $A$  and the **right-hand side vector**  $b$  are known.

$$\begin{bmatrix} 1 & x_0 & x_0^2 & \cdots & x_0^n \\ 1 & x_1 & x_1^2 & \cdots & x_1^n \\ 1 & x_2 & x_2^2 & \cdots & x_2^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^n \end{bmatrix}$$



Computed Tomography



# Overview (cont'd)

- The system may or may not have a solution, and if it has a solution, it may or may not be unique.
- Here we assume  $A$  is a square, **invertible** (**nonsingular**) matrix, so there exists a unique solution.
- In pure math, we get the solution

$$x^* = A^{-1}b$$

- But it is advisable to solve the system directly rather than explicitly computing the inverse.
- We will discuss two types of methods
  - Direct approach: **Gaussian elimination**
  - **Iterative approach** that generates  $x_1, x_2, \dots \rightarrow x^*$

# Example

$$\left\{ \begin{array}{rclclclcl} 6x_1 & - & 2x_2 & + & 2x_3 & + & 4x_4 & = & 16 \\ 12x_1 & - & 8x_2 & + & 6x_3 & + & 10x_4 & = & 26 \\ 3x_1 & - & 13x_2 & + & 9x_3 & + & 3x_4 & = & -19 \\ -6x_1 & + & 4x_2 & + & x_3 & - & 18x_4 & = & -34 \end{array} \right.$$

$$\left\{ \begin{array}{rclclclcl} 6x_1 & - & 2x_2 & + & 2x_3 & + & 4x_4 & = & 16 \\ & - & 4x_2 & + & 2x_3 & + & 2x_4 & = & -6 \\ & - & 12x_2 & + & 8x_3 & + & x_4 & = & -27 \\ & & 2x_2 & + & 3x_3 & - & 14x_4 & = & -18 \end{array} \right.$$

- The first equation was not altered, which is called the **pivot equation**.
- Keep going

$$\left\{ \begin{array}{rrrrrcl} 6x_1 & - & 2x_2 & + & 2x_3 & + & 4x_4 & = & 16 \\ & & - & 4x_2 & + & 2x_3 & + & 2x_4 & = & -6 \\ & & & & 2x_3 & - & 5x_4 & = & -9 \\ & & & & & - & 3x_4 & = & -3 \end{array} \right.$$

- The above system is said to be in the **upper triangular** form.
- All the linear systems are **equivalent**.

# Gaussian Elimination

- We just completed the first phase called **forward elimination**.
- We now proceed with the second phase: **back substitution**.
- Starting from ...

$$\left\{ \begin{array}{rclclclcl} 6x_1 & - & 2x_2 & + & 2x_3 & + & 4x_4 & = & 16 \\ & & - & 4x_2 & + & 2x_3 & + & 2x_4 & = & -6 \\ & & & & 2x_3 & - & 5x_4 & = & -9 \\ & & & & & & - & 3x_4 & = & -3 \end{array} \right.$$

# Generally

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & & \vdots \\ a_{i1} & a_{i2} & a_{i3} & \cdots & a_{in} \\ \vdots & \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_i \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ \vdots \\ b_i \\ \vdots \\ b_n \end{bmatrix}$$

multiplier

$$2 \leq i \leq n$$

$$\begin{cases} a_{ij} \leftarrow a_{ij} - \left( \frac{a_{i1}}{a_{11}} \right) a_{1j} & (1 \leq j \leq n) \\ b_i \leftarrow b_i - \left( \frac{a_{i1}}{a_{11}} \right) b_1 \end{cases}$$

# After the first step

Pivot element

Pivot equation  $\rightarrow$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ 0 & a_{22} & a_{23} & \cdots & a_{2n} \\ 0 & a_{23} & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & & \vdots \\ 0 & a_{i2} & a_{i3} & \cdots & a_{in} \\ \vdots & \vdots & \vdots & & \vdots \\ 0 & a_{n2} & a_{n3} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_i \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ \vdots \\ b_i \\ \vdots \\ b_n \end{bmatrix}$$

$$3 \leq i \leq n$$

$$\begin{cases} a_{ij} \leftarrow a_{ij} - \left( \frac{a_{i2}}{a_{22}} \right) a_{2j} & (2 \leq j \leq n) \\ b_i \leftarrow b_i - \left( \frac{a_{i2}}{a_{22}} \right) b_2 \end{cases}$$



# After the kth step

$$\begin{bmatrix}
 a_{11} & a_{12} & a_{13} & \cdots & \cdots & \cdots & a_{1n} \\
 0 & a_{22} & a_{23} & \cdots & \cdots & \cdots & a_{2n} \\
 0 & 0 & a_{33} & \cdots & \cdots & \cdots & a_{3n} \\
 \vdots & \vdots & \vdots & \ddots & & & \vdots \\
 0 & 0 & 0 & \cdots & \boxed{a_{kk}} & \cdots & a_{kn} \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
 0 & 0 & 0 & \cdots & a_{ik} & \cdots & a_{in} \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
 0 & 0 & 0 & \cdots & a_{nk} & \cdots & a_{nn}
 \end{bmatrix}
 \begin{bmatrix}
 x_1 \\
 x_2 \\
 x_3 \\
 \vdots \\
 x_k \\
 \vdots \\
 x_i \\
 \vdots \\
 x_n
 \end{bmatrix}
 =
 \begin{bmatrix}
 b_1 \\
 b_2 \\
 b_3 \\
 \vdots \\
 b_k \\
 \vdots \\
 b_i \\
 \vdots \\
 b_n
 \end{bmatrix}$$

$$k + 1 \leq i \leq n$$

$$\begin{cases}
 a_{ij} \leftarrow a_{ij} - \left( \frac{a_{ik}}{a_{kk}} \right) a_{kj} & (k \leq j \leq n) \\
 b_i \leftarrow b_i - \left( \frac{a_{ik}}{a_{kk}} \right) b_k
 \end{cases}$$

# Forward elimination

MatLab pseudo-code

# Back substitution

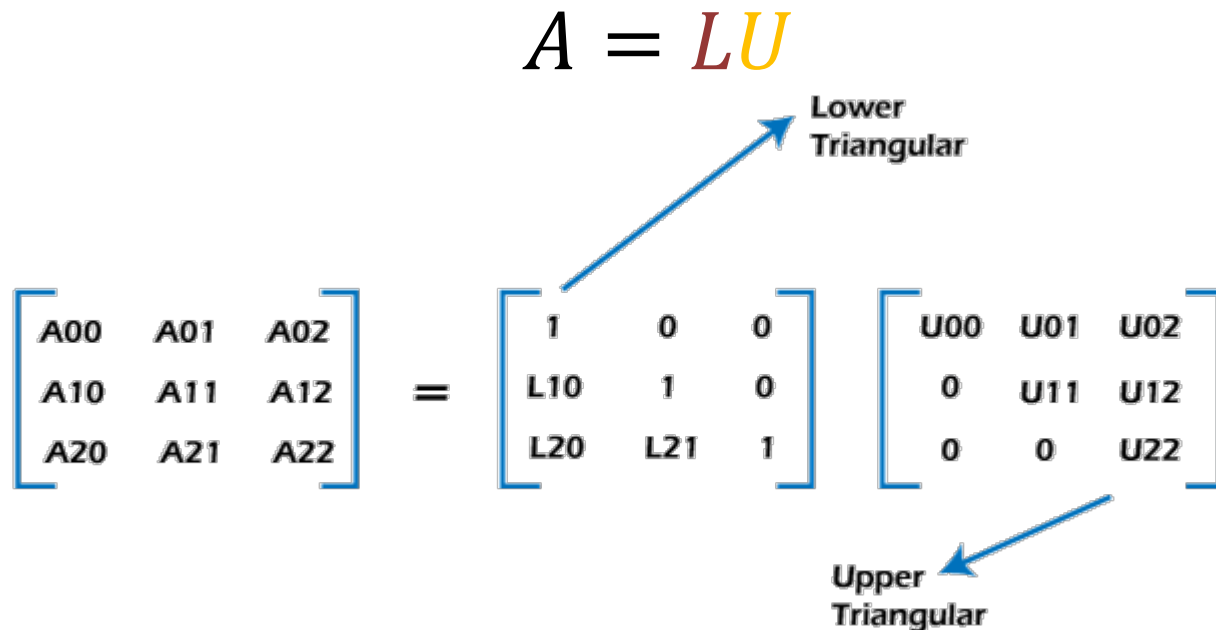
$$\left\{ \begin{array}{llll} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \cdots & & + a_{1n}x_n = b_1 \\ & a_{22}x_2 + a_{23}x_3 + \cdots & & + a_{2n}x_n = b_2 \\ & & a_{33}x_3 + \cdots & + a_{3n}x_n = b_3 \\ & & \ddots & \vdots \\ & & & \dots \vdots \\ & & a_{n-1,n-1}x_{n-1} & + a_{n-1,n}x_n = b_{n-1} \\ & & & a_{nn}x_n = b_n \end{array} \right.$$

Here coefficients are not the original ones, but are the ones that have been altered by the elimination.

# Mathematical analysis

# LU factorization

Gaussian elimination transforms a matrix into the product of a **unit lower triangular** matrix and an **upper triangular** matrix, i.e.,

$$A = LU$$


$$\begin{bmatrix} A_{00} & A_{01} & A_{02} \\ A_{10} & A_{11} & A_{12} \\ A_{20} & A_{21} & A_{22} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ L_{10} & 1 & 0 \\ L_{20} & L_{21} & 1 \end{bmatrix} \begin{bmatrix} U_{00} & U_{01} & U_{02} \\ 0 & U_{11} & U_{12} \\ 0 & 0 & U_{22} \end{bmatrix}$$

Lower  
Triangular

Upper  
Triangular

- Recall  $a_{ij} \leftarrow a_{ij} - \left(\frac{a_{ik}}{a_{kk}}\right) a_{kj}$
- We must expect all quantities to be infected with **roundoff errors**.
- The roundoff error in  $a_{kj}$  is multiplied by  $\left(\frac{a_{ik}}{a_{kk}}\right)$ .
- The small **pivot elements** would lead to large multipliers and to worse roundoff errors.
- Take-home message: Selecting a large value for pivoting every time (next lecture)

# Error analysis

For a linear system  $Ax = b$  having the **true solution**  $x$  and a **computed solution**  $\tilde{x}$ , we define

- Error vector:  $e = \tilde{x} - x$
- Residual vector:  $r = A\tilde{x} - b$

For two solutions, how do we evaluate which one is better?

- Look at the residual vector: smaller the better!