

# Lecture 6

## Gaussian Elimination

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# Gaussian Elimination

- Solving Triangular Systems
- Gaussian Elimination Without Pivoting
  - Hand Calculations
  - Cartoon Version
  - Algorithm
- Elementary Elimination Matrices And LU Factorization



# Gaussian Elimination

Gaussian elimination is a mostly general method for solving square systems.

We will work with systems in their matrix form, such as

$$\begin{aligned}x_1 + 3x_2 + 5x_3 &= 4 \\9x_1 + 7x_2 + 8x_3 &= 6 \\3x_1 + 2x_2 + 7x_3 &= 1,\end{aligned}$$

in its equivalent matrix form,

$$\begin{bmatrix} 1 & 3 & 5 \\ 9 & 7 & 8 \\ 3 & 2 & 7 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 6 \\ 1 \end{bmatrix}.$$



# Triangular Systems

The generic lower and upper triangular matrices are

$$L = \begin{bmatrix} l_{11} & 0 & \cdots & 0 \\ l_{21} & l_{22} & & 0 \\ \vdots & & \ddots & \vdots \\ l_{n1} & & \cdots & l_{nn} \end{bmatrix}$$

and

$$U = \begin{bmatrix} u_{11} & u_{12} & \cdots & u_{1n} \\ 0 & u_{22} & & u_{2n} \\ \vdots & & \ddots & \vdots \\ 0 & & \cdots & u_{nn} \end{bmatrix}$$

The triangular systems

$$Ly = b \qquad Ux = c$$

are easily solved by **forward substitution** and **backward substitution**, respectively



# Solving Triangular Systems

Solving for  $x_1, x_2, \dots, x_n$  for an upper triangular system is called **backward substitution**.

Listing 1: backward substitution (page 252)

```
1  given  $A$  (upper  $\triangle$ ),  $b$ 
2   $x_n = b_n/a_{nn}$ 
3  for  $i = n - 1 \dots 1$ 
4       $s = b_i$ 
5      for  $j = i + 1 \dots n$ 
6           $s = s - a_{i,j}x_j$ 
7      end
8       $x_i = s/a_{i,i}$ 
9  end
```



# Solving Triangular Systems

Solving for  $x_1, x_2, \dots, x_n$  for an upper triangular system is called **backward substitution**.

Listing 2: backward substitution (page 252)

```
1  given  $A$  (upper  $\triangle$ ),  $b$   
2   $x_n = b_n/a_{nn}$   
3  for  $i = n - 1 \dots 1$   
4       $s = b_i$   
5      for  $j = i + 1 \dots n$   
6           $s = s - a_{i,j}x_j$   
7      end  
8       $x_i = s/a_{i,i}$   
9  end
```

Using forward or backward substitution is sometimes referred to as performing a **triangular solve**.



# Operations?

cheap!

- begin in the bottom corner: 1 div
- row -2: 1 mult, 1 add, 1 div, or 3 FLOPS
- row -3: 2 mult, 2 add, 1 div, or 5 FLOPS
- row -4: 3 mult, 3 add, 1 div, or 7 FLOPS
- $\vdots$
- row -j: about  $2j - 1$  FLOPS

Total FLOPS?  $\sum_{j=1}^n 2j - 1 = 2\frac{n(n+1)}{2} - n$  or  $\mathcal{O}(n^2)$  FLOPS



# Gaussian Elimination

- Triangular systems are easy to solve in  $\mathcal{O}(n^2)$  FLOPS
- Goal is to transform an arbitrary, square system into an equivalent upper triangular system
- Then easily solve with backward substitution

This process is equivalent to the *formal solution* of  $Ax = b$ , where  $A$  is an  $n \times n$  matrix.

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



# Gaussian Elimination — Hand Calculations

Solve

$$x_1 + 3x_2 = 5$$

$$2x_1 + 4x_2 = 6$$

Subtract 2 times the first equation from the second equation

$$x_1 + 3x_2 = 5$$

$$-2x_2 = -4$$

This equation is now in triangular form, and can be solved by backward substitution.



# Gaussian Elimination — Hand Calculations

The elimination phase transforms the matrix and right hand side to an equivalent system

$$\begin{array}{rcl} x_1 + 3x_2 = 5 & \longrightarrow & x_1 + 3x_2 = 5 \\ 2x_1 + 4x_2 = 6 & & -2x_2 = -4 \end{array}$$

The two systems have the same solution. The right hand system is upper triangular.

Solve the second equation for  $x_2$

$$x_2 = \frac{-4}{-2} = 2$$

Substitute the newly found value of  $x_2$  into the first equation and solve for  $x_1$ .

$$x_1 = 5 - (3)(2) = -1$$



# Gaussian Elimination — Hand Calculations

When performing Gaussian Elimination by hand, we can avoid copying the  $x_i$  by using a shorthand notation.

For example, to solve:

$$A = \begin{bmatrix} -3 & 2 & -1 \\ 6 & -6 & 7 \\ 3 & -4 & 4 \end{bmatrix} \quad b = \begin{bmatrix} -1 \\ -7 \\ -6 \end{bmatrix}$$

Form the *augmented* system

$$\tilde{A} = [A \ b] = \left[ \begin{array}{ccc|c} -3 & 2 & -1 & -1 \\ 6 & -6 & 7 & -7 \\ 3 & -4 & 4 & -6 \end{array} \right]$$

The vertical bar inside the augmented matrix is just a reminder that the last column is the  $b$  vector.



# Gaussian Elimination — Hand Calculations

Add 2 times row 1 to row 2, and add (1 times) row 1 to row 3

$$\tilde{A}_{(1)} = \left[ \begin{array}{ccc|c} -3 & 2 & -1 & -1 \\ 0 & -2 & 5 & -9 \\ 0 & -2 & 3 & -7 \end{array} \right]$$

Subtract (1 times) row 2 from row 3

$$\tilde{A}_{(2)} = \left[ \begin{array}{ccc|c} -3 & 2 & -1 & -1 \\ 0 & -2 & 5 & -9 \\ 0 & 0 & -2 & 2 \end{array} \right]$$



# Gaussian Elimination — Hand Calculations

The transformed system is now in upper triangular form

$$\tilde{A}_{(2)} = \left[ \begin{array}{ccc|c} -3 & 2 & -1 & -1 \\ 0 & -2 & 5 & -9 \\ 0 & 0 & -2 & 2 \end{array} \right]$$

Solve by back substitution to get

$$x_3 = \frac{2}{-2} = -1$$

$$x_2 = \frac{1}{-2} (-9 - 5x_3) = 2$$

$$x_1 = \frac{1}{-3} (-1 - 2x_2 + x_3) = 2$$



# Gaussian Elimination — Cartoon Version

Start with the augmented system

$$\begin{bmatrix} x & x & x & x & x \\ x & x & x & x & x \\ x & x & x & x & x \\ x & x & x & x & x \end{bmatrix}$$

The  $x$ 's represent numbers, they are generally *not* the same values.

Begin elimination using the first row as the *pivot row* and the first element of the first row as the pivot element

$$\begin{bmatrix} \boxed{x} & x & x & x & x \\ x & x & x & x & x \\ x & x & x & x & x \\ x & x & x & x & x \end{bmatrix}$$



# Gaussian Elimination — Cartoon Version

- Eliminate elements under the pivot element in the first column.
- $x'$  indicates a value that has been changed once.

$$\begin{bmatrix} \boxed{x} & x & x & x & x \\ x & x & x & x & x \\ x & x & x & x & x \\ x & x & x & x & x \end{bmatrix} \longrightarrow \begin{bmatrix} \boxed{x} & x & x & x & x \\ 0 & x' & x' & x' & x' \\ x & x & x & x & x \\ x & x & x & x & x \end{bmatrix}$$

$$\longrightarrow \begin{bmatrix} \boxed{x} & x & x & x & x \\ 0 & x' & x' & x' & x' \\ 0 & x' & x' & x' & x' \\ x & x & x & x & x \end{bmatrix}$$

$$\longrightarrow \begin{bmatrix} \boxed{x} & x & x & x & x \\ 0 & x' & x' & x' & x' \\ 0 & x' & x' & x' & x' \\ 0 & x' & x' & x' & x' \end{bmatrix}$$

# Gaussian Elimination — Cartoon Version

- The pivot element is now the diagonal element in the second row.
- Eliminate elements under the pivot element in the second column.
- $x''$  indicates a value that has been changed twice.

$$\begin{bmatrix} x & x & x & x & x \\ 0 & \boxed{x'} & x' & x' & x' \\ 0 & x' & x' & x' & x' \\ 0 & x' & x' & x' & x' \end{bmatrix} \longrightarrow \begin{bmatrix} x & x & x & x & x \\ 0 & \boxed{x'} & x' & x' & x' \\ 0 & 0 & x'' & x'' & x'' \\ 0 & x' & x' & x' & x' \end{bmatrix}$$

$$\longrightarrow \begin{bmatrix} x & x & x & x & x \\ 0 & \boxed{x'} & x' & x' & x' \\ 0 & 0 & x'' & x'' & x'' \\ 0 & 0 & x'' & x'' & x'' \end{bmatrix}$$

# Gaussian Elimination — Cartoon Version

- The pivot element is now the diagonal element in the third row.
- Eliminate elements under the pivot element in the third column.
- $x'''$  indicates a value that has been changed three times.

$$\begin{bmatrix} x & x & x & x & x \\ 0 & x' & x' & x' & x' \\ 0 & 0 & \boxed{x''} & x'' & x'' \\ 0 & 0 & x'' & x'' & x'' \end{bmatrix} \longrightarrow \begin{bmatrix} x & x & x & x & x \\ 0 & x' & x' & x' & x' \\ 0 & 0 & \boxed{x''} & x'' & x'' \\ 0 & 0 & 0 & x''' & x''' \end{bmatrix}$$

# Gaussian Elimination — Cartoon Version

## Summary

- Gaussian Elimination is an orderly process for transforming an augmented matrix into an equivalent upper triangular form.
- The elimination operation at the  $k^{\text{th}}$  step is

$$\tilde{a}_{ij} = \tilde{a}_{ij} - (\tilde{a}_{ik}/\tilde{a}_{kk})\tilde{a}_{kj}, \quad i > k, \quad j \geq k$$

- Elimination requires three nested loops.
- The result of the elimination phase is represented by the image below.

$$\begin{bmatrix} x & x & x & x & x \\ x & x & x & x & x \\ x & x & x & x & x \\ x & x & x & x & x \end{bmatrix} \longrightarrow \begin{bmatrix} x & x & x & x & x \\ 0 & x' & x' & x' & x' \\ 0 & 0 & x'' & x'' & x'' \\ 0 & 0 & 0 & x''' & x''' \end{bmatrix}$$



# Gaussian Elimination

## Summary

- Transform a linear system into (upper) triangular form. i.e. transform lower triangular part to zero
- Transformation is done by taking linear combinations of rows
- Example:  $a = \begin{bmatrix} a_1 \\ a_2 \end{bmatrix}$
- If  $a_1 \neq 0$ , then

$$\begin{bmatrix} 1 & 0 \\ -a_2/a_1 & 1 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} a_1 \\ 0 \end{bmatrix}$$



# Gaussian Elimination Algorithm

## Listing 3: Forward Elimination beta

```
1  given  $A, b$ 
2
3  for  $k = 1 \dots n - 1$ 
4      for  $i = k + 1 \dots n$ 
5          for  $j = k \dots n$ 
6               $a_{ij} = a_{ij} - (a_{ik}/a_{kk})a_{kj}$ 
7          end
8       $b_i = b_i - (a_{ik}/a_{kk})b_k$ 
9  end
10 end
```

- the multiplier can be moved outside the  $j$ -loop
- no reason to actually compute 0

Challenge: The loops over  $i$  and  $j$  may be exchanged—why would one be preferable?

# Gaussian Elimination Algorithm

## Listing 4: Forward Elimination

```
1  given  $A, b$ 
2
3  for  $k = 1 \dots n - 1$ 
4      for  $i = k + 1 \dots n$ 
5           $xmult = a_{ik} / a_{kk}$ 
6           $a_{ik} = xmult$ 
7          for  $j = k + 1 \dots n$ 
8               $a_{ij} = a_{ij} - (xmult)a_{kj}$ 
9          end
10          $b_i = b_i - (xmult)b_k$ 
11     end
12 end
```



# Naive Gaussian Elimination Algorithm

- Forward Elimination
- + Backward substitution
- = Naive Gaussian Elimination



# Forward Elimination Cost?

What is the cost in converting from  $A$  to  $U$ ?

Step	Add	Multiply	Divide
1	$(n-1)^2$	$(n-1)^2$	$n-1$
2	$(n-2)^2$	$(n-2)^2$	$n-2$
$\vdots$			
n-1	1	1	1

or

add	$\sum_{j=1}^{n-1} j^2$
multiply	$\sum_{j=1}^{n-1} j^2$
divide	$\sum_{j=1}^{n-1} j$

# Forward Elimination Cost?

add	$\sum_{j=1}^{n-1} j^2$
multiply	$\sum_{j=1}^{n-1} j^2$
divide	$\sum_{j=1}^{n-1} j$

We know  $\sum_{j=1}^p j = \frac{p(p+1)}{2}$  and  $\sum_{j=1}^p j^2 = \frac{p(p+1)(2p+1)}{6}$ , so

add-subtracts	$\frac{n(n-1)(2n-1)}{6}$
multiply-divides	$\frac{n(n-1)(2n-1)}{6} + \frac{n(n-1)}{2} = \frac{n(n^2-1)}{3}$



# Forward Elimination Cost?

add-subtracts	$\frac{n(n-1)(2n-1)}{6}$
multiply-divides	$\frac{n(n^2-1)}{3}$
add-subtract for $b$	$\frac{n(n-1)}{2}$
multiply-divides for $b$	$\frac{n(n-1)}{2}$



# Back Substitution Cost

As before

add-subtract	$\frac{n(n-1)}{2}$
multiply-divides	$\frac{n(n+1)}{2}$

# Naive Gaussian Elimination Cost

Combining the cost of forward elimination and backward substitution gives

add-subtracts	$\frac{n(n-1)(2n-1)}{6} + \frac{n(n-1)}{2} + \frac{n(n-1)}{2}$ $= \frac{n(n-1)(2n+5)}{6}$
multiply-divides	$\frac{n(n^2-1)}{3} + \frac{n(n-1)}{2} + \frac{n(n+1)}{2}$ $= \frac{n(n^2+3n-1)}{3}$

So the total cost of add-subtract-multiply-divide is about

$$\frac{2}{3}n^3$$

$\Rightarrow$  double  $n$  results in a cost increase of a factor of 8



# Elimination Matrices

- Another way to zero out entries in a column of  $A$
- Annihilate entries below  $k^{\text{th}}$  element in  $a$  with matrix,  $M_k$ :

$$M_k a = \begin{bmatrix} 1 & \dots & 0 & 0 & \dots & 0 \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 1 & 0 & \dots & 0 \\ 0 & \dots & -m_{k+1} & 1 & \dots & 0 \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & -m_n & 0 & \dots & 1 \end{bmatrix} \begin{bmatrix} a_1 \\ \vdots \\ a_k \\ a_{k+1} \\ \vdots \\ a_n \end{bmatrix} = \begin{bmatrix} a_1 \\ \vdots \\ a_k \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

where  $m_i = a_i/a_k$ ,  $i = k+1, \dots, n$ .

- The divisor  $a_k$  is the “pivot” (and needs to be nonzero)



# Elimination Matrices

- Matrix  $M_k$  is an “elementary elimination matrix”
  - Adds a multiple of row  $k$  to each subsequent row, with “multipliers”  $m_i$
  - Result is zeros in the  $k^{\text{th}}$  column for rows  $i > k$ .
- $M_k$  is unit lower triangular and nonsingular
- $M_k = I - m_k e_k^T$  where  $m_k = [0, \dots, 0, m_{k+1}, \dots, m_n]^T$  and  $e_k$  is the  $k^{\text{th}}$  column of the identity matrix  $I$ .
- $M_k^{-1} = I + m_k e_k^T$ , which means  $M_k^{-1}$  is also lower triangular, and we will denote  $M_k^{-1} = L_k$ .

Can you prove  $M_k^{-1} = I + m_k e_k^T$ ?



# Elimination Matrices

- Suppose  $M_j$  and  $M_k$  are elementary elimination matrices with  $j > k$ , then

$$\begin{aligned}M_k M_j &= I - m_k e_k^T - m_j e_j^T + m_k e_k^T m_j e_j^T \\&= I - m_k e_k^T - m_j e_j^T + m_k (e_k^T m_j) e_j^T \\&= I - m_k e_k^T - m_j e_j^T\end{aligned}$$

because the  $k^{th}$  entry of vector  $m_j$  is zero (since  $j > k$ )

- Thus  $M_k M_j$  is essentially a union of their columns.
- Note this is also true for  $M_k^{-1} M_j^{-1}$ .



# Example

Let  $a = \begin{bmatrix} 2 \\ 4 \\ -2 \end{bmatrix}$ .

$$M_1 a = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 4 \\ -2 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix}$$

and

$$M_2 a = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1/2 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 4 \\ -2 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \\ 0 \end{bmatrix}$$



# Example

So

$$L_1 = M_1^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}, \quad L_2 = M_2^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1/2 & 1 \end{bmatrix}$$

which means

$$M_1 M_2 = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 1 & 1/2 & 1 \end{bmatrix}, \quad L_1 L_2 = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & -1/2 & 1 \end{bmatrix}$$



# Gaussian Elimination

- To reduce  $Ax = b$  to upper triangular form, first construct  $M_1$  with  $a_{11}$  as the pivot (eliminating the first column of  $A$  below the diagonal.)
- Then  $M_1Ax = M_1b$  still has the same solution.
- Next construct  $M_2$  with pivot  $a_{22}$  to eliminate the second column below the diagonal.
- Then  $M_2M_1Ax = M_2M_1b$  still has the same solution
- $M_{n-1} \dots M_1Ax = M_{n-1} \dots M_1b$
- Let  $M = M_nM_{n-1} \dots M_1$ . Then  $MAx = Mb$ , with  $MA$  upper triangular.
- Do back substitution on  $MAx = Mb$ .



# Another Way to Look at $A$

We've mentioned  $L$  and  $U$  today. Why?

Consider this

$$A = A$$

$$A = (M^{-1}M)A$$

$$A = (M_1^{-1}M_2^{-1} \dots M_n^{-1})(M_nM_{n-1} \dots M_1)A$$

$$A = (M_1^{-1}M_2^{-1} \dots M_n^{-1})((M_nM_{n-1} \dots M_1)A)$$

$$A = \qquad L \qquad U$$

But  $MA$  is upper triangular, and we've seen that  $M_1^{-1} \dots M_n^{-1}$  is lower triangular. Thus, we have an algorithm that factors  $A$  into two matrices  $L$  and  $U$ .



# Why is this “naive”?

## Example

$$A = \begin{bmatrix} 0 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$$

## Example

$$A = \begin{bmatrix} 1e-10 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$$

