# OF MINNESOTA TWIN CITIES



C S C I 5304

Fall 2021

#### COMPUTATIONAL ASPECTS OF MATRIX THEORY

Class time : MW 4:00 - 5:15 pm

Room: Keller 3-230 or Online

Instructor: Daniel Boley

Lecture notes:

http://www-users.cselabs.umn.edu/classes/Fall-2021/csci5304/

#### **SOLVING LINEAR SYSTEMS OF EQUATIONS**

- Background on linear systems
- Gaussian elimination and the Gauss-Jordan algorithms
- The LU factorization
- Gaussian Elimination with pivoting permutation matrices.
- Case of banded systems

## Background: Linear systems

**The Problem:** A is an  $n \times n$  matrix, and b a vector of  $\mathbb{R}^n$ . Find x such that:

$$Ax = b$$

 $igspace{}{x}$  is the unknown vector,  $oldsymbol{b}$  the right-hand side, and  $oldsymbol{A}$  is the coefficient matrix

## Example:

$$egin{cases} 2x_1 + 4x_2 + 4x_3 = 6 \ x_1 + 5x_2 + 6x_3 = 4 \ x_1 + 3x_2 + x_3 = 8 \end{cases} ext{ or } egin{pmatrix} 2 & 4 & 4 \ 1 & 5 & 6 \ 1 & 3 & 1 \end{pmatrix} egin{pmatrix} x_1 \ x_2 \ x_3 \end{pmatrix} = egin{pmatrix} 6 \ 4 \ 8 \end{pmatrix}$$

✓ Solution of above system?

Standard mathematical solution by Cramer's rule:

$$x_i = \det(A_i)/\det(A)$$

 $A_i = \text{matrix obtained by replacing } i\text{-th column by } b.$ 

Note: This formula is useless in practice beyond n=3 or n=4.

#### Three situations:

- 1. The matrix  ${m A}$  is nonsingular. There is a unique solution given by  ${m x} = {m A}^{-1}{m b}$ .
- 2. The matrix A is singular and  $b \in \text{Ran}(A)$ . There are infinitely many solutions.
- 3. The matrix A is singular and  $b \notin \operatorname{Ran}(A)$ . There are no solutions.

**Example:** (1) Let  $A = \begin{pmatrix} 2 & 0 \\ 0 & 4 \end{pmatrix}$   $b = \begin{pmatrix} 1 \\ 8 \end{pmatrix}$ . A is nonsingu-

lar  $\succ$  a unique solution  $x=\begin{pmatrix} 0.5\\2 \end{pmatrix}$  .

**Example:** (2) Case where A is singular &  $b \in \operatorname{Ran}(A)$ :

$$A=egin{pmatrix} 2 & 0 \ 0 & 0 \end{pmatrix}, \quad b=egin{pmatrix} 1 \ 0 \end{pmatrix}.$$

ightharpoonup infinitely many solutions:  $x(lpha)={0.5 \choose lpha} \ orall \ lpha.$ 

**Example:** (3) Let A same as above, but  $b = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ .

No solutions since 2nd equation cannot be satisfied

#### Triangular linear systems

#### Example:

$$egin{pmatrix} 2 & 4 & 4 \ 0 & 5 & -2 \ 0 & 0 & 2 \end{pmatrix} egin{pmatrix} x_1 \ x_2 \ x_3 \end{pmatrix} = egin{pmatrix} 2 \ 1 \ 4 \end{pmatrix}$$

- lacksquare One equation can be trivially solved: the last one.  $lacksquare x_3=2$
- $\succ x_3$  is known we can now solve the 2nd equation:

$$5x_2 - 2x_3 = 1 \rightarrow 5x_2 - 2 \times 2 = 1 \rightarrow x_2 = 1$$

ightharpoonup Finally  $oldsymbol{x}_1$  can be determined similarly:

$$2x_1 + 4x_2 + 4x_3 = 2 \rightarrow \dots \rightarrow x_1 = -5$$

## ALGORITHM: 1. Back-Substitution algorithm

```
For i=n:-1:1 do: t:=b_i For j=i+1:n do t:=t-a_{ij}x_j t:=b_i-(a_{i,i+1:n},x_{i+1:n}) t:=t-a_{ij}x_j t:=b_i-(a_{i,i+1:n},x_{i+1:n}) t:=t/a_{ii} End
```

- $\blacktriangleright$  We must require that each  $a_{ii} 
  eq 0$
- Operation count?

## Column version of back-substitution

#### Back-Substitution algorithm. Column version

```
For j=n:-1:1 do: x_j=b_j/a_{jj} For i=1:j-1 do b_i:=b_i-x_j*a_{ij} End
```

✓ Justify the above algorithm [Show that it does indeed compute the solution]

Analogous algorithms for lower triangular systems.

## Linear Systems of Equations: Gaussian Elimination

Back to arbitrary linear systems.

Principle of the method: Since triangular systems are easy to solve, we will transform a linear system into one that is triangular. Main operation: combine rows so that zeros appear in the required locations to make the system triangular.

Notation: use a Tableau:

$$\left\{egin{array}{lll} 2x_1+4x_2+4x_3=&2&&2&4&4&2\ x_1+3x_2+1x_3=&1& ext{tableau:}&1&3&1&1\ x_1+5x_2+6x_3=-6&&1&5&6&-6 \end{array}
ight.$$

Main operation used: scaling and adding rows.

**Example:** Replace row2 by: row2 -  $\frac{1}{2}$ \*row1:

This is equivalent to:

$$egin{bmatrix} 1 & 0 & 0 \ -rac{1}{2} & 1 & 0 \ 0 & 0 & 1 \end{bmatrix} imes egin{bmatrix} 2 & 4 & 4 & 2 \ 1 & 3 & 1 & 1 \ 1 & 5 & 6 & -6 \end{bmatrix} = egin{bmatrix} 2 & 4 & 4 & 2 \ 0 & 1 & -1 & 0 \ 1 & 5 & 6 & -6 \end{bmatrix}$$

The left-hand matrix is of the form

$$M = I - ve_1^T$$
 with  $v = egin{pmatrix} 0 \ rac{1}{2} \ 0 \end{pmatrix}$ 

## Linear Systems of Equations: Gaussian Elimination

Go back to original system. Step 1 must transform:

$$row_2 := row_2 - \frac{1}{2} \times row_1$$
:  $row_3 := row_3 - \frac{1}{2} \times row_1$ :

$$egin{array}{c|cccc} 2 & 4 & 4 & 2 \ 0 & 1 & -1 & 0 \ 1 & 5 & 6 & -6 \ \end{array}$$

> Equivalent to

$$[m{A},m{b}] 
ightarrow [m{M}_1m{A},m{M}_1m{b}]; \;\; m{M}_1 = m{I} - m{v}^{(1)}m{e}_1^T; \;\; m{v}^{(1)} = egin{pmatrix} 0 \ rac{1}{2} \ rac{1}{2} \end{pmatrix}$$

New system  $A_1x = b_1$ . Step 2 must now transform:

$$row_3 := row_3 - 3 imes row_2 : 
ightarrow egin{bmatrix} 2 & 4 & 4 & 2 \ 0 & 1 & -1 & 0 \ 0 & 0 & 7 & -7 \end{bmatrix}$$

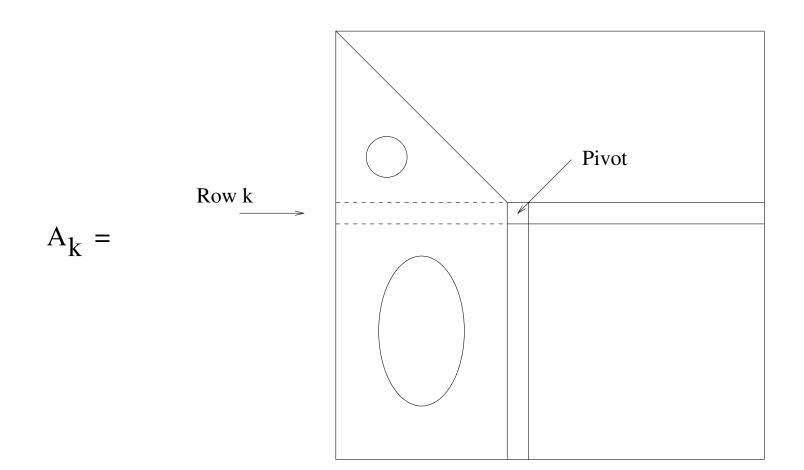
Equivalent to

$$egin{bmatrix} 1 & 0 & 0 \ 0 & 1 & 0 & imes \ 0 & -3 & 1 \ \end{bmatrix} imes egin{bmatrix} 2 & 4 & 4 & 2 \ 0 & 1 & -1 & 0 \ 0 & 3 & 4 & -7 \ \end{bmatrix} = egin{bmatrix} 2 & 4 & 4 & 2 \ 0 & 1 & -1 & 0 \ 0 & 0 & 7 & -7 \ \end{bmatrix}$$

Second transformation is as follows:

$$[A_1,b_1] 
ightarrow [M_2A_1,M_2b_1]; \; M_2=I-v^{(2)}e_2^T; \; v^{(2)}=egin{pmatrix} 0 \ 0 \ 3 \end{pmatrix}$$

Triangular system > Solve.



#### ALGORITHM: 2. Gaussian Elimination

```
1. For k = 1 : n - 1 Do:

2. For i = k + 1 : n Do:

3. piv := a_{ik}/a_{kk}

4. For j := k + 1 : n + 1 Do:

5. a_{ij} := a_{ij} - piv * a_{kj}

6. End

6. End

7. End
```

#### Operation count:

$$T = \sum_{k=1}^{n-1} \sum_{i=k+1}^{n} [1 + \sum_{j=k+1}^{n+1} 2] = \sum_{k=1}^{n-1} \sum_{i=k+1}^{n} (2(n-k) + 3) = ...$$

Complete the above calculation. Order of the cost?

#### $The \ LU \ factorization$

Now ignore the right-hand side from the transformations.

Observation: Gaussian elimination is equivalent to n-1 successive Gaussian transformations, i.e., multiplications with matrices of the form  $M_k=I-v^{(k)}e_k^T$ , where the first k components of  $v^{(k)}$  equal zero.

ightharpoonup Set  $A_0 \equiv A$ 

$$A o M_1 A_0 = A_1 o M_2 A_1 = A_2 o M_3 A_2 = A_3 \cdots \ o M_{n-1} A_{n-2} = A_{n-1} \equiv U$$

ightharpoonup Last  $A_k \equiv U$  is an upper triangular matrix.

 $_{\sf L}$  GvL 3. $\{1,3,5\}$  – Systems

lacksquare At each step we have:  $A_k = M_{k+1}^{-1} A_{k+1}$  . Therefore:

$$A_0 = M_1^{-1} A_1$$

$$= M_1^{-1} M_2^{-1} A_2$$

$$= M_1^{-1} M_2^{-1} M_3^{-1} A_3$$

$$= \dots$$

$$= M_1^{-1} M_2^{-1} M_3^{-1} \cdots M_{n-1}^{-1} A_{n-1}$$

- $L = M_1^{-1} M_2^{-1} M_3^{-1} \cdots M_{n-1}^{-1}$
- $\blacktriangleright$  Note: L is Lower triangular,  $A_{n-1}$  is upper triangular
- $\blacktriangleright$  LU decomposition : A=LU

#### How to get L?

$$L = M_1^{-1} M_2^{-1} M_3^{-1} \cdots M_{n-1}^{-1}$$

- Consider only the first 2 matrices in this product.
- $lacksquare ext{Note } M_k^{-1} = (I v^{(k)} e_k^T)^{-1} = (I + v^{(k)} e_k^T).$  So:

$$\boldsymbol{M}_{1}^{-1}\boldsymbol{M}_{2}^{-1} = (\boldsymbol{I} + \boldsymbol{v}^{(1)}\boldsymbol{e}_{1}^{T})(\boldsymbol{I} + \boldsymbol{v}^{(2)}\boldsymbol{e}_{2}^{T}) = \boldsymbol{I} + \boldsymbol{v}^{(1)}\boldsymbol{e}_{1}^{T} + \boldsymbol{v}^{(2)}\boldsymbol{e}_{2}^{T}.$$

Generally,

$$M_1^{-1}M_2^{-1}\cdots M_k^{-1} = I + v^{(1)}e_1^T + v^{(2)}e_2^T + \cdots v^{(k)}e_k^T$$

The L factor is a lower triangular matrix with ones on the diagonal. Column k of L, contains the multipliers  $l_{ik}$  used in the k-th step of Gaussian elimination.

A matrix  $oldsymbol{A}$  has an LU decomposition if

$$\det(A(1:k,1:k)) \neq 0$$
 for  $k = 1, \dots, n-1$ .

In this case, the determinant of A satisfies:

$$\det A = \det(U) = \prod_{i=1}^n u_{ii}$$

If, in addition,  $\boldsymbol{A}$  is nonsingular, then the LU factorization is unique.

Practical use: Show how to use the LU factorization to solve linear systems with the same matrix A and different b's.

LU factorization of the matrix  $m{A} = egin{pmatrix} 2 & 4 & 4 \\ 1 & 5 & 6 \\ 1 & 3 & 1 \end{pmatrix}$ ?

True or false: "Computing the LU factorization of matrix A involves more arithmetic operations than solving a linear system Ax = b by Gaussian elimination".

#### Gauss-Jordan Elimination

Principle of the method: We will now transform the system into one that is even easier to solve than triangular systems, namely a diagonal system. The method is very similar to Gaussian Elimination. It is just a bit more expensive.

Back to original system. Step 1 must transform:

 $row_2 := row_2 - 0.5 \times row_1$ :  $row_3 := row_3 - 0.5 \times row_1$ :

$$egin{array}{c|ccccc} 2 & 4 & 4 & 2 \ 0 & 1 & -1 & 0 \ 1 & 5 & 6 & -6 \ \hline \end{array}$$

 $row_1 := row_1 - 4 \times row_2$ :  $row_3 := row_3 - 3 \times row_2$ :

$$egin{array}{c|cccc} 2 & 0 & 8 & 2 \\ 0 & 1 & -1 & 0 \\ 0 & 3 & 4 & -7 \end{array}$$

#### There is now a third step:

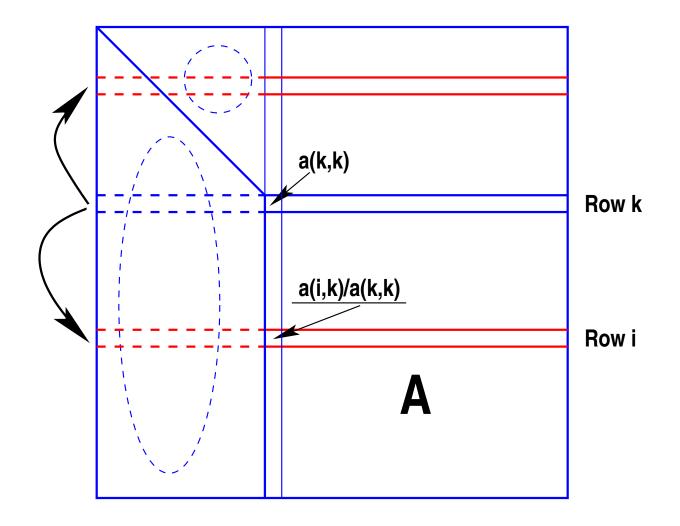
$\boldsymbol{x}$	0	0	$\boldsymbol{x}$
0	$\boldsymbol{x}$	0	$\boldsymbol{x}$
0	0	$\boldsymbol{x}$	$\boldsymbol{x}$

 $row_1 := row_1 - \frac{8}{7} \times row_3$ :  $row_2 := row_2 - \frac{-1}{7} \times row_3$ :

$$row_2 := row_2 - rac{-1}{7} imes row_3$$
:

$$egin{bmatrix} 2 & 0 & 0 & 10 \ 0 & 1 & 0 & 1 \ 0 & 0 & 7 & -7 \ \end{bmatrix}$$

Solution:  $x_3 = -1$ ;  $x_2 = -1$ ;  $x_1 = 5$ 



#### ALGORITHM: 3. Gauss-Jordan elimination

```
1. For k = 1 : n Do:

2. For i = 1 : n and if i! = k Do :

3. piv := a_{ik}/a_{kk}

4. For j := k + 1 : n + 1 Do :

5. a_{ij} := a_{ij} - piv * a_{kj}

6. End

6. End

7. End
```

### Operation count:

$$T = \sum_{k=1}^{n} \sum_{i=1}^{n-1} [1 + \sum_{j=k+1}^{n+1} 2] = \sum_{k=1}^{n} \sum_{i=1}^{n-1} (2(n-k) + 3) = \cdots$$

Complete the above calculation. Order of the cost? How does it compare with Gaussian Elimination?

```
function x = gaussj(A, b)
  function x = gaussj(A, b)
 solves A \times = b  by Gauss-Jordan  elimination
n = size(A,1);
A = [A,b];
for k=1:n
  for i=1:n
    if (i ~= k)
        piv = A(i,k) / A(k,k);
        A(i,k+1:n+1) = A(i,k+1:n+1) - piv*A(k,k+1:n+1);
     end
  end
end
x = A(:,n+1) ./ diag(A) ;
```

#### Gaussian Elimination: Partial Pivoting

Consider again Gaussian Elimination for the linear system

$$\left\{egin{array}{lll} 2x_1+2x_2+4x_3=&2&&2&4&2\ x_1+x_2+x_3=&1& ext{Or:}&1&1&1&1\ x_1+4x_2+6x_3=&-5&&1&4&6&-5 \end{array}
ight.$$

 $row_2 := row_2 - \frac{1}{2} \times row_1$ :  $row_3 := row_3 - \frac{1}{2} \times row_1$ :

$$egin{bmatrix} 2 & 2 & 4 & 2 \ 0 & 0 & -1 & 0 \ 1 & 4 & 6 & -5 \ \end{bmatrix}$$

 $\blacktriangleright$  Pivot  $a_{22}$  is zero. Solution : permute rows 2 and 3:

# Gaussian Elimination with Partial Pivoting

# Partial Pivoting

Row k

Largest | a ik | Pernuterous

General situation:

Always permute row  $m{k}$  with row  $m{l}$  such that

$$|a_{lk}| = \max_{i=k,\dots,n} |a_{ik}|$$

More 'stable' algorithm.

```
function x = gaussp(A, b)
   function x = guassp(A, b)
   solves A \times = b by Gaussian elimination with
  partial pivoting/
n = size(A,1);
 A = [A,b]
 for k=1:n-1
     [t, ip] = \max(abs(A(k:n,k)));
     ip = ip+k-1;
%% swap
     temp = A(k,k:n+1) ;
     A(k,k:n+1) = A(ip,k:n+1);
     A(ip,k:n+1) = temp;
     for i=k+1:n
     piv = A(i,k) / A(k,k);
     \bar{A}(i,k+1:n+1) = A(i,k+1:n+1) - piv*A(k,k+1:n+1);
   end
 end
 x = backsolv(A,A(:,n+1));
                                          GvL 3.{1,3,5} – Systems
3-28
```

## Pivoting and permutation matrices

- A permutation matrix is a matrix obtained from the identity matrix by permuting its rows
- $\succ$  For example for the permutation  $\pi=\{3,1,4,2\}$  we obtain

$$P = egin{pmatrix} 0 & 0 & 1 & 0 \ 1 & 0 & 0 & 0 \ 0 & 0 & 0 & 1 \ 0 & 1 & 0 & 0 \end{pmatrix}$$

Important observation: the matrix PA is obtained from A by permuting its rows with the permutation  $\pi$ 

$$(PA)_{i,:}=A_{\pi(i),:}$$

$$P = egin{pmatrix} 0 & 0 & 1 & 0 \ 1 & 0 & 0 & 0 \ 0 & 0 & 0 & 1 \ 0 & 1 & 0 & 0 \end{pmatrix} \;\; A = egin{pmatrix} 1 & 2 & 3 & 4 \ 5 & 6 & 7 & 8 \ 9 & 0 & -1 & 2 \ -3 & 4 & -5 & 6 \end{pmatrix} \; ?$$

- Any permutation matrix is the product of interchange permutations, which only swap two rows of I.
- lacksquare Notation:  $oldsymbol{E}_{ij}=$  Identity with rows i and j swapped

**Example:** To obtain  $\pi=\{3,1,4,2\}$  from  $\pi=\{1,2,3,4\}$  – we need to swap  $\pi(2)\leftrightarrow\pi(3)$  then  $\pi(3)\leftrightarrow\pi(4)$  and finally  $\pi(1)\leftrightarrow\pi(2)$ . Hence:

$$P = egin{pmatrix} 0 & 0 & 1 & 0 \ 1 & 0 & 0 & 0 \ 0 & 0 & 0 & 1 \ 0 & 1 & 0 & 0 \end{pmatrix} = E_{1,2} imes E_{3,4} imes E_{2,3}$$

∠
10 In the previous example where

$$>> A = [1234; 5678; 90-12; -34-56]$$

Matlab gives det(A) = -896. What is det(PA)?

At each step of G.E. with partial pivoting:

$$M_{k+1}E_{k+1}A_k = A_{k+1}$$

where  $E_{k+1}$  encodes a swap of row k+1 with row l>k+1.

Notes: (1)  $E_i^{-1}=E_i$  and (2)  $M_j^{-1} imes E_{k+1}=E_{k+1} imes ilde{M}_j^{-1}$  for  $k\geq j$ , where  $ilde{M}_j$  has a permuted Gauss vector:

$$egin{align} (I + v^{(j)} e_j^T) E_{k+1} &= E_{k+1} (I + E_{k+1} v^{(j)} e_j^T) \ &\equiv E_{k+1} (I + ilde{v}^{(j)} e_j^T) \ &\equiv E_{k+1} ilde{M}_j \ \end{gathered}$$

Here we have used the fact that above row k+1, the permutation matrix  $E_{k+1}$  looks just like an identity matrix.

#### Result:

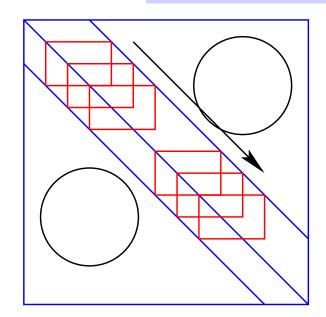
$$egin{aligned} A_0 &= E_1 M_1^{-1} A_1 \ &= E_1 M_1^{-1} E_2 M_2^{-1} A_2 = E_1 E_2 ilde{M}_1^{-1} M_2^{-1} A_2 \ &= E_1 E_2 ilde{M}_1^{-1} M_2^{-1} E_3 M_3^{-1} A_3 \ &= E_1 E_2 E_3 ilde{M}_1^{-1} ilde{M}_2^{-1} M_3^{-1} A_3 \ &= \dots \ &= E_1 \cdots E_{n-1} \ imes ilde{M}_1^{-1} ilde{M}_2^{-1} ilde{M}_3^{-1} \cdots ilde{M}_{n-1}^{-1} \ imes A_{n-1} \end{aligned}$$

#### In the end

$$PA = LU$$
 with  $P = E_{n-1} \cdots E_1$ 

## Special case of banded matrices

- Banded matrices arise in many applications
- lacksquare A has upper bandwidth q if  $a_{ij}=0$  for j-i>q
- lacksquare A has lower bandwidth p if  $|a_{ij}=0|$  for i-j>p



ightharpoonup Simplest case: tridiagonal ightharpoonup p=q=1.

First observation: Gaussian elimination (no pivoting) preserves the initial banded form. Consider first step of Gaussian elimination:

```
2. For i = 2:n Do:

3. a_{i1} := a_{i1}/a_{11} (pivots)

4. For j := 2:n Do:

5. a_{ij} := a_{ij} - a_{i1} * a_{1j}

6. End

7. End
```

If A has upper bandwidth q and lower bandwidth p then so is the resulting [L/U] matrix.  $\blacktriangleright$  Band form is preserved (induction)

△ 11 Operation count?

#### What happens when partial pivoting is used?

If A has lower bandwidth p, upper bandwidth q, and if Gaussian elimination with partial pivoting is used, then the resulting U has upper bandwidth p+q. L has at most p+1 nonzero elements per column (bandedness is lost).

ightharpoonup Simplest case: tridiagonal ightharpoonup p=q=1.

# Example:

$$A = egin{pmatrix} 1 & 1 & 0 & 0 & 0 \ 2 & 1 & 1 & 0 & 0 \ 0 & 2 & 1 & 1 & 0 \ 0 & 0 & 2 & 1 & 1 \ 0 & 0 & 0 & 2 & 1 \end{pmatrix}$$