

## Solving Linear Systems

### IE170: Algorithms in Systems Engineering: Lecture 30

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- Last time we learned about how to solve systems  $Ax = b$ , when  $A$  was symmetric and positive-definite.
- The key was to factor the matrix into two triangular matrices  $A = LU$
- In the case that  $A$  is SPD, then we can always do this, and in fact  $U = L^T$ .
- What if  $A$  is *not* SPD?
- The workhorse in this case is the LU-decomposition
- LU-decomposition is very related to (the well-known) Gaussian elimination, a fact we will try to make clear today...



## Gaussian Elimination

- An example for today. Let's solve it...

$$\begin{aligned}x_1 + x_2 + 2x_3 &= 3 \\2x_1 + 3x_2 + x_3 &= 2 \\3x_1 - x_2 - x_3 &= 6\end{aligned}$$

- Subtract twice first equation from the second
- Subtract 3 times the first equation from the third
- Then add 4 times second equation to the third
- You've made a triangular system!
- What were the matrices that produce this?



## Elementary Dear Watson!



- We reduced the columns by taking **linear combinations** of the rows of the matrix.
- This implies that the reduction process can be thought of as a multiplication of  $A$  **on the left** by some matrix
- What does the matrix look like?
- It is an **elementary matrix** of the form

$$E = I - uv^T$$

- In fact, it's a special form of an elementary matrix: It will be a **unit lower triangular** matrix with multipliers only in one column



## The Elimination Matrix

- Let's find a matrix  $M_1$  that reduces the first column of  $A$ .

$$M_1 = \begin{pmatrix} 1 & & & & & \\ -m_{21} & 1 & & & & \\ -m_{31} & & 1 & & & \\ \vdots & \vdots & & \ddots & & \\ -m_{n1} & 0 & 0 & \cdots & 1 & \end{pmatrix}$$

- By our properties of matrix multiplication, this matrix
  - Leaves the first row of  $A$  alone
  - Takes  $-m_{21}$  times the first row, adds the second row, and puts this in the second row of the new matrix  $M_1A$
  - Takes  $-m_{31}$  times the first row of  $A$ , adds the third row, and puts this in the third row of the new matrix  $M_1A$
  - (And So On...)



## The UpShot

- To eliminate the first column, we want

$$m_{i1} = \frac{a_{i1}}{a_{11}}$$

- Note:** These were **exactly** the multipliers we used in our simple example
- $a_{11}$  is called the **pivot element**, and this reduction only works if the pivot element is  $\neq 0$
- Next time: What happens if pivot element is 0 (or small)



## Let's Carry On

$$M_1A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ 0 & a_{22}^{(2)} & \cdots & a_{2n}^{(2)} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & a_{n2}^{(2)} & \cdots & a_{nn}^{(2)} \end{pmatrix}$$

- To reduce the second column, we would like a matrix

$$M_1 = \begin{pmatrix} 1 & & & & \\ 0 & 1 & & & \\ 0 & -m_{32} & 1 & & \\ \vdots & \vdots & & \ddots & \\ 0 & -m_{n2} & 0 & \cdots & 1 \end{pmatrix}$$



## Matrix Effect

- Again, the matrix  $M_2$  will...
  - Leave first row of  $M_1A$  unchanged in  $M_2M_1A$
  - Leave second row of  $M_1A$  unchanged in  $M_2M_1A$
  - Take  $-m_{32}$  times second row + third row in  $M_2M_1A$

### Lather Rinse Repeat

- Repeat this  $n - 1$  times
- In the end, we get  $M_{n-1} \cdots M_2M_1A = U$
- Fact:** The product of unit lower triangular matrices is unit lower triangular
- So in the end we have  $MA = U$ , with  $M$  unit lower triangular, and  $U$  upper triangular
- This process is known as **Gaussian Elimination**, and the matrix  $M$  is known as the **product form of the LU factorization**



## Finding LU Directly

- Here is a recursive method for finding the LU factorization
- We'll divide the matrix  $A$  into four pieces:

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \quad (1)$$

$$= \begin{bmatrix} a_{11} & w^T \\ v & A' \end{bmatrix} \quad (2)$$

- Next, we'll use **row operations** to change  $v$  into the zero vector and record the operations in another matrix.



## Finding the LU Decomposition (cont.)

- By simple multiplication, you can verify the following factorization of  $A$ :

$$A = \begin{bmatrix} a_{11} & w^T \\ v & A' \end{bmatrix} \quad (3)$$

$$= \begin{bmatrix} 1 & 0 \\ v/a_{11} & I \end{bmatrix} \begin{bmatrix} a_{11} & w^T \\ 0 & A' - vw^T/a_{11} \end{bmatrix} \quad (4)$$

- We can show that if  $A$  is nonsingular, then so is  $A' - vw^T/a_{11}$ .
- So we can recursively call the method to factor the  $(n-1) \times (n-1)$  matrix  $A' - vw^T/a_{11}$ .
- Applying this recursion  $n$  times yields the desired factorization



## Finding the LU Decomposition (cont.)

- To see how to get the factorization from the recursive application of the algorithm, we have the following.

$$A = \begin{bmatrix} 1 & 0 \\ v/a_{11} & I \end{bmatrix} \begin{bmatrix} a_{11} & w^T \\ 0 & A' - vw^T/a_{11} \end{bmatrix} \quad (5)$$

$$= \begin{bmatrix} 1 & 0 \\ v/a_{11} & I \end{bmatrix} \begin{bmatrix} a_{11} & w^T \\ 0 & L'U' \end{bmatrix} \quad (6)$$

$$= \begin{bmatrix} 1 & 0 \\ v/a_{11} & L' \end{bmatrix} \begin{bmatrix} a_{11} & w^T \\ 0 & U' \end{bmatrix} \quad (7)$$

- This shows how to obtain the factorization recursively.
- This can also be done iteratively and "in place."



## The Algorithm

LU-DECOMPOSITION( $A$ )

```

1   $n \leftarrow \text{rows}[L]$ 
2  for  $k \leftarrow 1$  to  $n$ 
3  do
4       $u_{kk} \leftarrow a_{kk}$ 
5      for  $i \leftarrow 1$  to  $n$ 
6      do
7           $\ell_{ik} \leftarrow a_{ik}/u_{kk}$ 
8           $u_{ki} \leftarrow a_{ki}$ 
9      for  $i \leftarrow k+1$  to  $n$ 
10     do
11         for  $j \leftarrow k+1$  to  $n$ 
12         do
13              $a_{ij} \leftarrow a_{ij} - \ell_{ik}u_{kj}$ 

```



## LU $\approx$ Gaussian Elimination

- We either have  $A = LU$  or we have  $MA = U$
- $M$  is unit lower triangular, and in fact the inverse of a unit lower triangular matrix is unit lower-triangular, so  $A = M^{-1}U$ , and since the elements of  $L$  and  $U$  are unique, it must be that  $L = M^{-1}$
- Because of the special structure of  $M$ , we have a (fairly) remarkable relationship



## LU $\approx$ Gaussian Elimination

- The relationship is the following:

$$M^{-1} = (M_{n-1} \cdots M_2 M_1)^{-1} = M_1^{-1} M_2^{-1} \cdots = L$$

$$L = \begin{pmatrix} 1 & & & & \\ m_{21} & 1 & & & \\ m_{31} & m_{32} & 1 & & \\ \vdots & & & \ddots & \\ m_{n1} & m_{n2} & m_{n3} & \cdots & 1 \end{pmatrix}$$

where the  $m_{ik}$  are **the multipliers from Gaussian elimination!**

- So  $L$  and  $U$  can be derived directly from the elimination process:

$$\ell_{ik} = m_{ik} = \frac{a_{ik}^{(k)}}{a_{kk}^{(k)}} \quad u_{kj} = a_{kj}^{(k)}$$

