

2/21/2019

LU Factorization

Last time: We saw by example that

we can take  $A \in \mathbb{R}^{n \times n}$  and reduce it to an upper triangular matrix  $U$  with a sequence of matrices  $L_1, L_2, \dots, L_N$

of the form  $I + \mu_{rs} E^{(rs)}$ ,

where  $E^{(rs)}$  is the matrix whose only nonzero entry is in the  $r$ th row and  $s$ th column:

i.e. if  $E^{(rs)} \in \mathbb{R}^{3 \times 3}$ , then

$$\mu_{31} E^{(31)} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ \mu_{31} & 0 & 0 \end{pmatrix}.$$

Also, we saw that if

$$L = I + \mu_{rs} E^{(rs)},$$

then  $L^{-1} = I - \mu_{rs} E^{(rs)}$ .

Thm: 2.1 These are true for any integer  $n \geq 2$ .

(i) The product of two lower  $\Delta$  matrices of order  $n$  is lower  $\Delta$  of order  $n$ .

(ii) The product of two unit lower  $\Delta$  matrices of order  $n$  is unit lower  $\Delta$  of order  $n$ .

(iii) a lower  $\Delta$  matrix is nonsingular  $\Leftrightarrow$  all the diagonal elements are nonzero.

(iv) The inverse of a nonsingular lower  $\Delta$  matrix of order  $n$  is lower  $\Delta$  of order  $n$ .

(v) The inverse of a unit lower  $\Delta$  matrix of order  $n$  is unit lower  $\Delta$  of order  $n$ .

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Pf of iv (By induction).

Base case:  $n=2$ .

Let  $L = \begin{pmatrix} a & 0 \\ c & b \end{pmatrix}$ . Note that we can write down  $L^{-1}$  explicitly:

$$L^{-1} = \frac{1}{\det L} \begin{pmatrix} b & 0 \\ -c & a \end{pmatrix}.$$

Since  $\det L = ab$ , we are implicitly assuming  $a \neq 0$  and  $b \neq 0$ . Thus,  $L^{-1}$  is lower  $\Delta$ .

Inductive hypothesis Assume true for  $2 \leq n \leq k$ .

i.e. the inverse of a nonsingular lower  $\Delta$  matrix of order  $k$  is lower  $\Delta$  of order  $k$ .

Prove true for  $k+1$ . Let  $L$  be a

lower  $\Delta$  triangular matrix of order  $k+1$ .

We can partition  $L$  and its inverse  $L^{-1}$  as:

$$L = \begin{pmatrix} L_1 & \underline{0} \\ \underline{r}^T & \alpha \end{pmatrix}, \quad L^{-1} = \begin{pmatrix} X & \underline{y} \\ \underline{z}^T & \beta \end{pmatrix}.$$

By definition,  $LL^{-1} = \underline{I}_{k+1}$ , so

$$L_1 X = \underline{I}_k, \quad L_1 \underline{y} = \underline{0}, \quad \underline{r}^T X + \alpha \underline{z}^T = \underline{0}^T$$

$$\text{and } \underline{r}^T \underline{y} + \alpha \beta = 1.$$

By induction, since  $X = L_1^{-1}$ ,  $X$  is lower  $\Delta$ . Also, since  $L_1$  is invertible,  $\underline{y} = \underline{0}$ . This shows  $L^{-1}$  is lower  $\Delta$ .

## Computing LU.

How do we actually compute the LU Factorization?

If we assume that  $A$  can be factored as  $A = LU$ ,  $L$  unit lower  $\Delta$  and  $U$  upper  $\Delta$ , then express:

$$A = (a_{ij}), \quad L = (l_{ij}), \quad U = (u_{ij}).$$

Then by definition of matrix multiplication

$$a_{ij} = \sum_{k=1}^n l_{ik} u_{kj}, \quad 1 \leq i, j \leq n.$$

Since  $L$  and  $U$  are lower and upper triangular, the sum only goes to  $\min(i, j)$

$$a_{ij} = \sum_{k=1}^{\min(i, j)} l_{ik} u_{kj}.$$

let's break into cases:

Case 1:  $1 \leq j < i \leq n$

$$a_{ij} = \sum_{k=1}^j l_{ik} u_{kj}$$

Case 2:  $1 \leq i \leq j \leq n$

$$a_{ij} = \sum_{k=1}^i l_{ik} u_{kj}$$

look at case 1:

$$a_{ij} = \sum_{k=1}^j l_{ik} u_{kj} = \sum_{k=1}^{j-1} l_{ik} u_{kj} + l_{ij} u_{jj}$$

Solving for  $l_{ij}$ , we get

$$l_{ij} = \frac{1}{u_{jj}} \left( a_{ij} - \sum_{k=1}^{j-1} l_{ik} u_{kj} \right)$$

$$i = 2, \dots, n.$$

$$j = 1, \dots, i-1.$$

look at case 2: and use  $l_{jj} = 1, 1 \leq j \leq n$ .

$$a_{ij} = \sum_{k=1}^i l_{ik} u_{kj} = \sum_{k=1}^{i-1} l_{ik} u_{kj} + l_{ii} u_{ij}$$

$$\Rightarrow u_{ij} = a_{ij} - \sum_{k=1}^{i-1} l_{ik} u_{kj}$$
$$i = 1, \dots, n$$
$$j = i, \dots, n$$

Example: Consider the matrix:

$$A = \begin{bmatrix} 6 & 2 & 3 \\ 2 & 2 & 0 \\ 3 & 0 & 3 \end{bmatrix}.$$

Note that because  $L = \begin{bmatrix} 1 & & 0 \\ & 1 & \\ 0 & & 1 \end{bmatrix}$

we can immediately read off the 1st row of U:

$$u_{11} = a_{11} = 6, \quad u_{12} = a_{12} = 2, \quad u_{13} = a_{13} = 3.$$

1st row of L:  $l_{11} = 1.$

2nd row of L:

$$l_{21} = \frac{1}{u_{11}} \left( a_{21} - \sum_{k=1}^0 l_{2k} u_{k1} \right) = \frac{a_{21}}{u_{11}} = \frac{2}{6} = \frac{1}{3}.$$

$$l_{22} = 1.$$

2nd row of U:

$$\begin{aligned} u_{22} &= a_{22} - \sum_{k=1}^1 l_{2k} u_{k2} = a_{22} - l_{21} u_{12} \\ &= 2 - \left(\frac{1}{3}\right)(2) = 1\frac{1}{3}. \end{aligned}$$

$$u_{23} = a_{23} - \sum_{k=1}^1 l_{2k} u_{k3} = a_{23} - l_{21} u_{13} = 0 - \left(\frac{1}{3}\right)(3) = -1.$$

3rd row of L : 0

$$l_{31} = \frac{1}{u_{11}} \left( a_{31} - \sum_{k=1}^1 l_{3k} u_{k1} \right) = \frac{a_{31}}{u_{11}} = \frac{3}{6} = \frac{1}{2}.$$

$$l_{32} = \frac{1}{u_{22}} \left( a_{32} - \sum_{k=1}^1 l_{3k} u_{k2} \right)$$

$$= \frac{1}{u_{22}} l_{31} u_{12} = \frac{-3}{4} \left( \frac{1}{2} \right) (2) = -\frac{3}{4}.$$

$$l_{33} = 1.$$

3rd row of U :

$$u_{33} = a_{33} - \sum_{k=1}^2 l_{3k} u_{k3}$$

$$= a_{33} - l_{31} u_{13} - l_{32} u_{23}$$

$$= 3 - \left( \frac{1}{2} \right) (3) - \left( -\frac{3}{4} \right) (-1)$$

$$= \frac{3}{2} - \frac{3}{4} = \frac{3}{4}.$$

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What about the process for computing LU breaking down? i.e.  $u_{jj} = 0$  for some  $j$ .

We need a definition:

Thm

Thm 2.2:  $A \in \mathbb{R}^{n \times n}$ ,  $n \geq 2$ .

Suppose every leading principle submatrix  $A^{(k)} \in \mathbb{R}^{k \times k}$ , with  $1 \leq k < n$ , is nonsingular. ( $A$  is not required to be nonsingular).

Then  $A = LU$ , where  $L$  is unit lower  $\Delta$  and  $U$  is upper  $\Delta$ .

Definition 2.4:  $A \in \mathbb{R}^{n \times n}$ ,  $n \geq 2$ .  
 $1 \leq k \leq n$ .

The leading principle submatrix of order  $k$  of  $A$  is defined as  $A^{(k)} \in \mathbb{R}^{k \times k}$  whose element in row  $i$  and column  $j$  is equal to the element of matrix  $A$  in row  $i$  and column  $j$ ,  $1 \leq i, j \leq k$ .

Pf: (of Thm 2.2)

We do induction on  $n$ .

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Base case  $n=2$

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}, \quad a \neq 0 \text{ by assumption.}$$

we want to find  $L = \begin{pmatrix} 1 & 0 \\ m & 1 \end{pmatrix}$ ,  $U = \begin{pmatrix} u & v \\ 0 & \eta \end{pmatrix}$

Satisfying: 
$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ m & 1 \end{pmatrix} \begin{pmatrix} u & v \\ 0 & \eta \end{pmatrix}.$$

We have the following equations

$$a = u.$$

$$b = v$$

$$mu = c$$

$$mv + \eta = d.$$

Since  $a \neq 0$ , we can solve:

$$m = \frac{c}{u} = \frac{c}{a}.$$

$$\eta = d - mv = d - \left(\frac{c}{a}\right) b.$$

So the base case  $n=2$  holds.

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Inductive hypothesis:

Statement is true for matrices of order  $k$ ,  
 $2 \leq k < n$ .

Prove: Statement is true for matrices of  
order  $k+1$ . let  $A \in \mathbb{R}^{(k+1) \times (k+1)}$ .

Partition  $A$  as follows:

$$A = \begin{pmatrix} A^{(k)} & b \\ c^T & d \end{pmatrix}$$

We can apply the inductive hypothesis  
on  $A^{(k)}$ , i.e.  $\exists L^{(k)}, U^{(k)}$  so that

$$A^{(k)} = L^{(k)} U^{(k)}$$

Guess  $L, U$  look like

$$L = \begin{pmatrix} L^{(k)} & 0 \\ m^T & 1 \end{pmatrix}, \quad U = \begin{pmatrix} U^{(k)} & v \\ 0^T & \gamma \end{pmatrix}$$

If  $A = LU$ , we get four  
equalities:

$$L^{(k)} U^{(k)} = A^{(k)}, \quad L^{(k)} v = b, \quad m^T U^{(k)} = c^T, \quad m^T v + \gamma = d.$$

$L^{(k)} v = b$  allows us to uniquely determine  $v$  since  $L^{(k)}$  is invertible.

Also, note that

$$\begin{aligned}\det(A^{(k)}) &= \det(L^{(k)} U^{(k)}) \\ &= \det(L^{(k)}) \det(U^{(k)}) \\ &= 1 * \det(U^{(k)}).\end{aligned}$$

Since  $A^{(k)}$  is invertible,  $\det(U^{(k)}) \neq 0$ , and so  $U^{(k)}$  is invertible. Thus

$$m^T U^{(k)} = c^T$$

uniquely determines  $m$ .

Then we can use the last equation

$$m^T v + \eta = d \quad \text{to get } \eta, \text{ given } m \text{ and } v.$$

<sup>random</sup>  
Some <sup>^</sup> comments: about "LU-like" Factorizations

Remark: In our first example, we reduced the system of equations as:

$$\underbrace{\begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}}_A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 1 \\ 0 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

Equivalently, we multiplied  $A$  by  $\begin{pmatrix} 1 & 0 \\ -1 & 1 \end{pmatrix}$

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

So we have  $L_1 A = U_1$ , where

$$L_1 = \begin{pmatrix} 1 & 0 \\ -1 & 1 \end{pmatrix}, \quad U_1 = \begin{pmatrix} 1 & 1 \\ 0 & -2 \end{pmatrix}$$

Now, we can write  $A = L_1^{-1} U_1$ ,  
with  $L_1^{-1} = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix}$ .

$$A = L_1^{-1} U_1 = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & -2 \end{pmatrix},$$

This is an LU factorization for  $A$ ,

but it is not unique! In particular,  
we can squeeze in  $\begin{pmatrix} a & 0 \\ 0 & a \end{pmatrix} \begin{pmatrix} \frac{1}{a} & 0 \\ 0 & \frac{1}{a} \end{pmatrix}$

For  $a \neq 0$  to get

$$\begin{aligned} A &= \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & -2 \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} a & 0 \\ 0 & a \end{pmatrix} \begin{pmatrix} \frac{1}{a} & 0 \\ 0 & \frac{1}{a} \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & -2 \end{pmatrix} \\ &= \begin{pmatrix} a & 0 \\ a & a \end{pmatrix} \begin{pmatrix} \frac{1}{a} & \frac{1}{a} \\ 0 & -\frac{2}{a} \end{pmatrix}. \end{aligned}$$

Fact: (I think, please check!). An LU  
Factorization is unique if  $L$  is required  
to be unit lower  $\Delta$ .

Base on the nonuniqueness of a factorization of  $A$  into ~~upper~~ a product of upper  $\Delta$  and lower  $\Delta$  matrices, we explicitly define the (unique) LU Factorization of  $A$  as Follows:

Definition The LU Factorization of  $A$ ,

If it exists, is the product  $A = LU$ , where  $L$  is unit lower  $\Delta$  and  $U$  is upper  $\Delta$ .

Remk: let's say we are given an arbitrary matrix  $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$  and we want to transform this to upper  $\Delta$ , using the techniques just described. We can multiply the 1st row by  $-\frac{c}{a}$  and add to the second row:

$$\begin{pmatrix} 1 & 0 \\ -\frac{c}{a} & 1 \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} a & b \\ 0 & -\frac{c}{a}(b+d) \end{pmatrix}.$$

But, what if  $a \approx 0$ . This is not stable. (we will define "not stable" later in the class).

It suffices to say that we need to be careful about ~~dividing~~ dividing by numbers close to zero when we do Gaussian elimination.

Example:  $A = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}$ .



If we try to use (blindly) our process to convert  $A$  to an upper  $\Delta$  matrix, we are in trouble since  $a_{11} = 0$ .

But, we can multiply  $A$  by a permutation matrix  $P$  which rearranges

the rows.

$$P = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}.$$

$$\begin{aligned} PA &= \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix} \\ &= \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}. \end{aligned}$$

Trivially, we have  $PA = LU$ ,

where  $L = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$ ,  $U = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ .

Thm: let  $n \geq 2$  and  $A \in \mathbb{R}^{n \times n}$ . There exists a permutation matrix  $P$ , a unit lower  $\Delta$  matrix  $L$ , and an upper  $\Delta$  matrix  $U$ , so that  $P, L, U \in \mathbb{R}^{n \times n}$  and

$$PA = LU.$$