

# Math 449: Numerical Applied Mathematics

## Lecture 09

09/20/2017

Today's topic: LU Decomposition

Today: LU decomposition

Last time: Gaussian Elimination  $A\vec{x} = \vec{b}$

1. Reduce system to upper triangular form
2. Solve upper triangular system by back substitution.

$$\begin{pmatrix} u_{11} & \dots & u_{1n} \\ & & \vdots \\ & & u_{nn} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix}$$

$$x_i = \frac{1}{u_{ii}} \left( b_i - \sum_{j=i+1}^n u_{ij} x_j \right) \quad \text{for } i = 2, \dots, n \quad n^2 \text{ total operations}$$

$u_{ii} \neq 0$

1st idea: Divide-and-Conquer

Suppose  $n$  is even,  $AB = C$ , split  $A, B, C$  into  $\frac{n}{2} \times \frac{n}{2}$  blocks

$$\begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix} = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}$$

We can multiply blockwise just like  $2 \times 2$  matrices.

$$C_{11} = A_{11} B_{11} + A_{12} B_{21}$$

$$C_{21} = A_{21} B_{11} + A_{22} B_{21}$$

$$C_{12} = A_{11} B_{12} + A_{12} B_{22}$$

$$C_{22} = A_{21} B_{12} + A_{22} B_{22}$$

Reduce  $n \times n$  matrix multiplication to

$$8 \quad \frac{n}{2} \times \frac{n}{2} \quad \text{matrix mults}$$

$$4 \quad \frac{n}{2} \times \frac{n}{2} \quad \text{adds}$$

$$4 \left( \frac{n}{2} \right)^2 = n^2 \quad \text{ordering adds}$$

$T(n)$ : # of ops to multiply 2  $n \times n$  matrices.

For simplicity, assume  $n$  is a power of 2.

$$\text{Recurrence relation} \quad \begin{cases} T(n) = 8T\left(\frac{n}{2}\right) + n^2 \\ T(1) = 1 \end{cases}$$

Special case of the master theorem Lemma  $\begin{cases} T(n) = aT\left(\frac{n}{b}\right) + cn^d \\ T(1) = 1 \end{cases}$

if  $\log_b a > d$ , then  $T(n) = O(n^{\log_b a})$

Proof Sketch

$$\begin{aligned}
 T(n) &= aT\left(\frac{n}{b}\right) + cn^d \\
 &= a\left(aT\left(\frac{n}{b^2}\right) + c\left(\frac{n}{b}\right)^d\right) + cn^d \\
 &= a^2T\left(\frac{n}{b^2}\right) + cn^d\left(1 + \frac{a}{b^d}\right) \\
 &= a^3T\left(\frac{n}{b^3}\right) + cn^d\left(1 + \frac{a}{b^d} + \left(\frac{a}{b^d}\right)^2\right) \\
 &= a^{\log_b n} T(1) + cn^d\left(1 + \frac{a}{b^d} + \dots + \left(\frac{a}{b^d}\right)^{\log_b n - 1}\right)
 \end{aligned}$$

1st term

$$a^{\log_b n} = n^{\log_b a}$$

2nd term

$$O\left(n^d \left(\frac{a}{b^d}\right)^{\log_b n}\right)$$

$$n^d \left(\frac{a}{b^d}\right)^{\log_b n} = n^d \frac{a^{\log_b n}}{n^d} = n^{\log_b a}$$

□

Divide + Conquer Algorithm

$$\log_b a = \log_2 8 = 3 > 2 = d \quad O(n^3)$$

Gauss Algorithm for multiplying complex number

$$\begin{aligned}
 (a+ib)(c+id) &= (ac+bd) + i(ad+bc) \\
 &= [(a+b)c - b(c+d)] + i[a(d-c) + (a+b)c]
 \end{aligned}$$

Note: very useful for  $(A+iB)(C+iD)$

Strassen's Algo:

$$\begin{aligned}
 M_1 &= (A_{11} + A_{22})(B_{11} + B_{22}) \\
 M_2 &= (A_{21} + A_{22})B_{11} \\
 M_3 &= A_{11}(B_{12} - B_{22}) \\
 M_4 &= A_{22}(B_{21} - B_{11}) \\
 M_5 &= (A_{11} + A_{12})B_{22} \\
 M_6 &= (A_{21} - A_{11})(B_{11} + B_{12}) \\
 M_7 &= (A_{12} - A_{22})(B_{21} + B_{22})
 \end{aligned}$$

$$C_{11} = M_1 + M_4 - M_5 + M_7$$

$$C_{12} = M_3 + M_5$$

$$C_{21} = M_2 + M_4$$

$$C_{22} = M_1 - M_2 + M_3 + M_6$$

7 matrix mult

18 adds/subs

$$T(n) = 7T\left(\frac{n}{2}\right) + 18\left(\frac{n}{2}\right)^2 \quad \frac{9}{2}n^2$$

$$T(1) = 1$$

$$\log_2 7 \approx 2.8074 > 2 \Rightarrow \text{Strassen is } O(n^{\log_2 7}) \approx O(n^{2.807})$$

$$\text{Current Record} \approx O(n^{2.3728639})$$

F. Le Gall 2014.

Conj.  $O(n^{2+\epsilon})$  for any  $\epsilon > 0$

## Fast Matrix Multiplication

$$\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix} \quad 2 \times 2 + 2 \times 2 = 8$$

Strassen 1969

## Cost of Gaussian Elimination

$$\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \longrightarrow \begin{pmatrix} a_{11} & a_{12} \\ 0 & a_{22} - \frac{a_{12} \cdot a_{21}}{a_{11}} \end{pmatrix}$$

unit - lower triangular matrix.

$$L = \begin{pmatrix} 1 & 0 \\ \frac{a_{21}}{a_{11}} & 1 \end{pmatrix} \quad L^{-1} = \begin{pmatrix} 1 & 0 \\ -\frac{a_{21}}{a_{11}} & 1 \end{pmatrix}$$

$$U = L^{-1}A$$

$$A = LU$$

more generally,  $n \times n$  matrix

$$U = L_{(N)} \cdots L_{(1)} A \quad \text{where each } L_{(i)} \text{ eliminates 1 entry below diagonal}$$

$$L_{(i)} = \begin{pmatrix} 1 & & & \\ & \ddots & & \\ & & 1 & \\ & & & \ddots & \\ & & & & 1 \end{pmatrix}$$

$$A = (L_N \cdots L_1)^{-1} \cdot U = L \cdot U$$

Note: LU decomposition only works if we never encounter a 0 on the diagonal.

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

$$\begin{pmatrix} 1 & 0 \\ l_{21} & 1 \end{pmatrix} \begin{pmatrix} u_{11} & u_{12} \\ 0 & u_{22} \end{pmatrix} = \begin{pmatrix} u_{11} & u_{12} \\ l_{21}u_{11} & l_{21}u_{12} + u_{22} \end{pmatrix}$$

$$\text{let } u_{11} = 0 \quad u_{12} = 1 \quad \text{but then } l_{21} \cdot u_{11} = 0 \neq 1$$

Once we have  $A = LU$ , it's easy to solve  $A\vec{x} = \vec{b}$

$$\underbrace{LU}_{\vec{y}} \vec{x} = \vec{b}$$

Solve  $L\vec{y} = \vec{b}$  by substitution

$U\vec{x} = \vec{y}$  by back-substitution

$O(n^2)$

Reminder: not every matrix has an LU decomp.