

# Honors Numerical Analysis

Lecture 8

## Numerical Linear Algebra

In infinite precision - exact - arithmetic, solving linear systems ~~is~~ comes down to a question of whether or not the system is consistent (or not).

On the computer, however, the situation is different due to round-off error and finite precision calculations.

First, a few comments:

Block matrices: 
$$A = \begin{pmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & \dots \\ \vdots & & \end{pmatrix}$$

is notation where each  $A_{ij}$  is a matrix.

$$A = (\vec{a}_1 \ \vec{a}_2 \ \dots \ \vec{a}_n)$$

↑  
vectors.

In Julia,  $\vec{a}_j$  can be accessed by  $A[:,j]$ .

A sub-matrix can be extracted as  $A[i_1:i_2, j_1:j_2]$ .

Using Linear Algebra

`a = rand(5,4)`  
`opnorm(a, p)`

(induced  
p norm)

also `p=Inf` is option

Note: at the

Julia prompt,

type "?" for help,  
search functions

①

# Gaussian Elimination

We all know how Gaussian elimination works -

Recall: Let  $A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$

$A\vec{x} = \vec{b}$  can be solved by eliminating  $a_{21}$ :

$$\left( \begin{array}{cc|c} a_{11} & a_{12} & b_1 \\ a_{21} & a_{22} & b_2 \end{array} \right) \sim \left( \begin{array}{cc|c} a_{11} & a_{12} & b_1 \\ 0 & a_{22} - \frac{a_{12}a_{21}}{a_{11}} & b_2 - \frac{a_{21}a_{12}}{a_{11}} \end{array} \right)$$

This implies  $x_2 = \frac{b_2 - \frac{a_{21}a_{12}}{a_{11}}}{a_{22} - \frac{a_{12}a_{21}}{a_{11}}}$  and then

$$x_1 = \frac{1}{a_{11}} (b_1 - a_{12}x_2)$$

This is back-substitution.

But - the process can easily fail: Ex:  $a_{11} = 0$  or

$$a_{22} - \frac{a_{21}a_{12}}{a_{11}} = 0$$

A successful algorithm would need to involve pivoting rows.

(will return to this later,)

How expensive is G.E.?

Count flops: "floating point operations" - a single  $+$ ,  $-$ ,  $\times$ ,  $\div$

To put  $A$  in echelon form: 
$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ 0 & \tilde{a}_{22} & \tilde{a}_{23} & \dots \\ 0 & 0 & \dots & \dots \end{pmatrix}$$

Consider the algorithm:

Loop over columns  $j=1, \dots, n-1$

Loop over rows  $i=j+1, \dots, n$

① Compute  $a_{ij}/a_{jj}$  (1 flop)

② Compute  $\text{row } i - \frac{a_{ij}}{a_{jj}} \text{row } j$  ( $2(n-j)$  flops)

③ Compute  $b_i - \frac{a_{ij}}{a_{jj}} b_j$  (2 flops)

The total is obtained by summing:

$$\sum_{j=1}^{n-1} \sum_{i=j+1}^n (2(n-j) + 3) = \sum_{j=1}^{n-1} (n-j)(2(n-j) + 3)$$

$$\approx \theta(n^3) \text{ flops (work out details at home.)}$$

(see Driscoll & Braun for asymptotic notation, flop counting.)

Another way to think about Gaussian Elimination:

### LU Factorization

Each row operation corresponds to multiplication by an "elementary" lower triangular matrix on the left.

Ex:  $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 0 \end{pmatrix}$

$$\underbrace{\begin{pmatrix} 1 & 0 & 0 \\ -4 & 1 & 0 \\ -7 & 0 & 1 \end{pmatrix}}_{L_1} \underbrace{\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 0 \end{pmatrix}}_A = \begin{pmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \\ 0 & -6 & -21 \end{pmatrix}$$

$$\underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -2 & 1 \end{pmatrix}}_{L_2} L_1 A = \underbrace{\begin{pmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \\ 0 & 0 & -9 \end{pmatrix}}_U$$

So  $L_2 L_1 A = U$

$\Rightarrow A = \underbrace{L_1^{-1} L_2^{-1}}_{\text{also lower triangular}} U$

Then to solve  $A \vec{x} = \vec{b}$   
 $\underbrace{L_2 L_1}_{LU} \vec{x} = \vec{b}$

Solve  $L_2 \vec{y} = \vec{b}$  FWD substitution (\*)

then  $U \vec{x} = \vec{y}$  BKWD substitution (\*\*)

(\*) and (\*\*) only cost  $\mathcal{O}(n^2)$  flops

## Pivoting

Of course G.E. or LU may fail if there is a zero or small number in a pivot position.

Solution is to interchange rows, resulting in a factorization of the form  $PA = LU$   
↑ permutation matrix.

Example Let  $\epsilon$  be a small number, e.g.  $\epsilon < \epsilon_{\text{mach}}$   
↑ machine precision

Then consider 
$$\begin{pmatrix} \epsilon & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

Exact solution is:  $x_1 = 1 + \frac{\epsilon}{\epsilon-1} \approx 1$

$$x_2 = 1 - \frac{\epsilon}{\epsilon+1} \approx 1$$

Without pivoting, in floating point arithmetic, G.E.

gives us:

$$\left( \begin{array}{cc|c} \epsilon & 1 & 1 \\ 1 & 1 & 2 \end{array} \right) \sim \left( \begin{array}{cc|c} \epsilon & 1 & 1 \\ 0 & -\frac{1}{\epsilon} & -\frac{1}{\epsilon} \end{array} \right)$$

Can be alleviated totally by pivoting for next time...

$$\Rightarrow \begin{aligned} x_2 &= 1 \\ x_1 &= \frac{1-1}{\epsilon} = 0 \quad \left( \text{or maybe } \frac{\epsilon_{\text{mach}}}{\epsilon} \right) \end{aligned}$$

The wrong answer!