

Example The ill-conditioned Hilbert matrices, defined by  $h_{ij} = \frac{1}{(i+j-1)}$

for example  $H_4 = \begin{bmatrix} 1 & \frac{1}{2} & \frac{1}{3} & \frac{1}{4} \\ \frac{1}{2} & \frac{1}{3} & \frac{1}{4} & \frac{1}{5} \\ \frac{1}{3} & \frac{1}{4} & \frac{1}{5} & \frac{1}{6} \\ \frac{1}{4} & \frac{1}{5} & \frac{1}{6} & \frac{1}{7} \end{bmatrix}$

See if you can find  $L$  and  $U$ .

$$U = \begin{bmatrix} 1 & \frac{1}{2} & \frac{1}{3} & \frac{1}{4} \\ 0 & \frac{1}{2} & \frac{1}{2} & \frac{3}{10} \\ 0 & 0 & \frac{1}{10} & \frac{9}{100} \\ 0 & 0 & 0 & -\frac{1}{1200} \end{bmatrix}$$

↑  
Last row could look like a row of zeros if we only carry 2 decimal places

(3) Given that  $A = \begin{pmatrix} 1 & -2 & -3 \\ 0 & 1 & 2 \\ -1 & -2 & 1 \end{pmatrix}$ ,  $B = \begin{pmatrix} -1 & 0 & 4 \\ -2 & 1 & 2 \\ 0 & -2 & 1 \end{pmatrix}$ ,  $C = \begin{pmatrix} -1 & 1 & 1 \\ -2 & 1 & 2 \\ 1 & -2 & 1 \end{pmatrix}$ ,  $D =$

$E = \begin{pmatrix} -2 & 1 & 4 \\ -2 & 1 & 1 \\ 1 & -2 & 1 \end{pmatrix}$ . Answer the following questions:

- (a) Multiply  $\begin{pmatrix} A & 0 \\ 0 & B \end{pmatrix} \begin{pmatrix} C & 0 \\ 0 & D \end{pmatrix}$  in an efficient manner
- (b) Multiply  $\begin{pmatrix} A^T & 0 \\ 0 & B \end{pmatrix} \begin{pmatrix} C^T & 0 \\ 0 & D \end{pmatrix}$  in an efficient manner
- (c) Multiply  $\begin{pmatrix} A & 0 \\ 0 & C \end{pmatrix} \begin{pmatrix} B & 0 \\ 0 & D \end{pmatrix}$  in an efficient manner
- (d) Multiply  $\begin{pmatrix} A & 0 \\ 0 & B^T \end{pmatrix} \begin{pmatrix} D & 0 \\ 0 & C \end{pmatrix}$  in an efficient manner

(a)  $\begin{pmatrix} AC & 0 \\ 0 & BD \end{pmatrix}$

$AC =$

$BD =$

- (5) If we derive an algorithm that is  $\mathcal{O}(n^3)$  that solves the system  $Ax = b$ . How would the length of time (approximately) to compute  $x$  be effected if instead of a 100 by 100 matrix  $\bar{A}$  we had a 300 by 300 matrix  $A$ ?

$A$  8 times as large as  $\bar{A}$

$(3)^3 = 27$

(9) Given  $A = \begin{pmatrix} -1 & 1 \\ 2 & 3 \end{pmatrix}$  and  $\kappa_p(A)$  is the condition number using the  $p$ -norm. Find the following.

(a)  $\kappa_1(A)$

(b)  $\kappa_2(A)$

(c)  $\kappa_\infty(A)$

$$A = \begin{pmatrix} -1 & 1 \\ 2 & 3 \end{pmatrix}$$

$$A = \begin{pmatrix} -\frac{3}{5} & \frac{1}{5} \\ \frac{2}{5} & \frac{1}{5} \end{pmatrix}$$

$$\kappa_2(A) = \|A\|_2 \|A^{-1}\|_2$$

$$\|x\|_2 = \sqrt{x_1^2 + x_2^2 + x_3^2 + \dots + x_n^2}$$

How do we translate into matrix norm?

$$\sqrt{(-1)^2 + (1)^2} = \sqrt{2}$$

$$\sqrt{2^2 + 3^2} = \sqrt{13}$$

$$\sqrt{2 + 13} = \sqrt{15}$$

$$A^{-1} = \begin{pmatrix} -\frac{3}{5} & \frac{1}{5} \\ \frac{2}{5} & \frac{1}{5} \end{pmatrix}$$

$$\sqrt{\frac{9+1}{25}} = \frac{\sqrt{10}}{5}$$

$$\frac{\sqrt{5}}{5}$$

$$\sqrt{\frac{10}{25} + \frac{5}{25}} = \frac{\sqrt{15}}{5}$$

$$\sqrt{\left(\sqrt{a_{11}^2 + a_{12}^2}\right)^2 + \left(\sqrt{a_{21}^2 + a_{22}^2}\right)^2}$$

$$= \sqrt{a_{11}^2 + a_{12}^2 + a_{21}^2 + a_{22}^2}$$

$$\|A\|_2 \|A^{-1}\|_2 = (\sqrt{15}) \left(\frac{\sqrt{15}}{5}\right) = \frac{15}{5} = \boxed{3}$$

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} a_1 & b_1 \\ c_1 & d_1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$a a_1 + b c_1 = 1$$

$$a b_1 + b d_1 = 0$$

$$c a_1 + d c_1 = 0$$

$$c b_1 + d d_1 = 1$$

$$-a_1 + c_1 = 1$$

$$-b_1 + d_1 = 0$$

$$2a_1 + 3c_1 = 0$$

$$\Rightarrow 2b_1 + 3d_1 = 1$$

$$5b_1 = 1$$

$$b_1 = \frac{1}{5}$$

$$c_1 = 1 + a_1$$

$$2a_1 + 3(1 + a_1) = 0$$

$$5a_1 = -3$$

$$a_1 = -\frac{3}{5}$$

$$c_1 = \frac{2}{5}$$

$$\|A\|_2 \|A^{-1}\|_2 = (\sqrt{15}) \left(\frac{1}{\sqrt{5}}\right) = \sqrt{3}$$

## LU - Decomposition

\* - relatively easy

\* - challenging

\* - average

### ① Overall Code Structure

- Use comments to show the location of different segments of the code.

### ② Check for more efficient algorithm that use Gaussian based on the structure of matrix (sparse, banded)

↳ up to 2 people working together

### ③ Check to see if matrix meets criteria for a Cholesky decomposition

### ④ Programme in Cholesky decomposition

### ⑤ Calculate condition number for the matrix A be able to use any p-norm. You may not use the built in norms/condition #s in NumPy. (2 to 3 people)

### ⑥ Determine pivots, keep track of row swap

### ⑦ Compute column $i$ for matrix $L$

### ⑧ Compute row $i$ for matrix $U$

### ⑨ Write code for back substitution

### ⑩ Write code for forward substitution

### ⑪ State $L, P, U$

### ⑫ State solution to the system and condition number

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State solution to the system  
and condition number  
(print out information,

Communication or Discard server

due March 19<sup>th</sup>

Let's have assignments decided by next Tuesday  
March 5<sup>th</sup>,

## Error Analysis of Gaussian Elimination

- Round off error
- Backward stability

Why small pivots should be avoided

Example Consider the linear system

$$\begin{bmatrix} 0.002 & 1.231 & 2.471 \\ 1.196 & 3.165 & 2.543 \\ 1.475 & 4.271 & 2.142 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3.704 \\ 6.904 \\ 7.888 \end{bmatrix}$$

This system is well conditioned  
 $\kappa_2(A) \approx 30$

Let's say we perform Gaussian Elimination  
without pivoting.  
Then

$$l_{21} = \frac{1.196}{0.002} = 598$$
$$l_{31} = \frac{1.475}{0.002} = 737.5$$

These values are multiplied by the first row and subtracted  
from their respective rows

$$3.165 - 598(1.231) = -732.9$$

$$\rightarrow 3.165 - 598(1.231) = -732.9$$

The digits in 3.165 were lost due to roundoff

This type of information loss is called swamping

This occurs at other entries of matrix as well.

$$\left[ \begin{array}{ccc|c} 0.002 & 1.231 & 2.471 & \\ \hline 598 & -732.9 & -1475 & \\ 737.5 & -903.6 & -1820 & \end{array} \right]$$

$$l_{32} = \frac{-903.6}{-732.9} = \underline{1.233}$$

$$-1820 - (1.233)(-1475) = -1$$

$$\left[ \begin{array}{ccc|c} 0.002 & 1.231 & 2.471 & \\ \hline 598 & -732.9 & -1475 & \\ 737.5 & 1.233 & -1 & \end{array} \right]$$

$$L = \begin{bmatrix} 1 & 0 & 0 \\ 598 & 1 & 0 \\ 737.5 & 1.233 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 3.704 \\ 6.904 \\ 7.858 \end{bmatrix} \quad Ly = b$$

$$\begin{aligned} y_1 &= 3.704 \\ y_2 &= -2208 \\ y_3 &= -2.000 \end{aligned}$$

$$\begin{aligned} 598y_1 + y_2 &= 6.904 \\ y_2 &= 6.904 - 598(3.704) \end{aligned}$$

$$\begin{aligned} Ux &= y \\ x_3 &= \frac{y_3}{u_{33}} = \frac{-2.000}{-1.000} \\ &= +2 \end{aligned}$$

$$\text{We can show } x = \begin{bmatrix} 4.000 \\ -1.012 \\ 2.000 \end{bmatrix}$$

But the actual solution is

$$x = \begin{bmatrix} 1.000 \\ 1.000 \\ 1.000 \end{bmatrix}$$

But the actual  $\begin{matrix} \text{L} & \text{U} \end{matrix}$  solution is  $x = \begin{bmatrix} 1.000 \\ 1.000 \\ 1.000 \end{bmatrix}$

This can be found by storing only 4 digits per entry, but allowing for partial pivoting.

## Backward Error Analysis of Gaussian Elimination

Gaussian elimination involves accumulation of sums

$$\sum_{j=1}^n w_j$$

there are many ways to add  $n$  numbers together.

We could work left to right or we could reorder the terms of the summation,

Because of floating point arithmetic the order in which we add could play a significant role.

Goal If a sum is accumulated in floating-point arithmetic, the computation is always backwards stable, regardless of the way the sum is accumulated,

(let  $u$  be the unit roundoff.  
 $O(u^2)$  order of  $u^2$ .)

For example,  $(1 + \alpha_1)(1 + \alpha_2) = (1 + \beta)$  where  $\beta = \alpha_1 + \alpha_2 + \alpha_1\alpha_2$   
If  $|\alpha_1| \approx u$   $|\alpha_2| \approx u$  then  $\alpha_1\alpha_2 = O(u^2)$ .

so  $\beta = \alpha_1 + \alpha_2 + O(u^2)$

proposition Suppose we compute the sum  $\sum_{j=1}^n w_j$  using floating point arithmetic with unit roundoff  $u$ .

using floating point arithmetic with unit roundoff  $u$ .

Then 
$$f\left(\sum_{j=1}^n w_j\right) = \sum_{j=1}^n w_j(1 + \delta_j)$$

where  $|\delta_j| \leq (n-1)u + O(u^2)$ , regardless of the order in which the terms are added,

$\delta_j$  can be termed relative backwards errors

The inequality  $|\delta_j| \leq (n-1)u + O(u^2)$  overestimates  $|\delta_j|$

$n-1$  reflects the fact that there are at most  $n-1$  additions

thus most  $\delta_j$  are sums of many fewer than  $n-1$  roundoffs.

Bounds like  $|\delta_j| \leq (n-1)u + O(u^2)$  takes into account the worst possible cases

$\delta_j$  is more likely to be closer to  $u$  than  $(n-1)u$ ,

the factor  $(n-1)$  is ignored in practice