

Numerical Linear Algebra

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Well posed problem, ill conditioned problem, condition Number

- ► Recap of Previous Lecture
- III conditioned linear system and matrix properties
- Condition number of a matrix
- Properties of Cond(A)
- Perturbations in right hand side
- Perturbations in coefficients
- Perturbations in right hand side and coefficients
- ► Q & A

Recap of Previous Lecture

- Computational project 1
- Convergence of matrix sequences
- Matrix Series
- Triangular systems of linear equations
- Condition number

$$\begin{pmatrix} 1 & -1 & \dots & \dots & -1 \\ 0 & 1 & \dots & \dots & -1 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 1 & -1 \\ 0 & 0 & \dots & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_{n-1} \\ x_n \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \\ \dots \\ -1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 & -1 & \dots & \dots & -1 & | & -1 \\ 0 & 1 & \dots & \dots & -1 & | & -1 \\ \dots & \dots & \dots & \dots & | & \dots \\ 0 & 0 & \dots & 1 & -1 & | & -1 \\ 0 & 0 & \dots & 0 & 1 & | & 1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & -1 & \dots & \dots & -1 \\ 0 & 1 & \dots & \dots & -1 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 1 & -1 \\ 0 & 0 & \dots & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_{n-1} \\ x_n \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \\ \dots \\ -1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 & -1 & \dots & \dots & -1 & | & -1 \\ 0 & 1 & \dots & \dots & -1 & | & -1 \\ \dots & \dots & \dots & \dots & | & \dots \\ 0 & 0 & \dots & 1 & -1 & | & -1 \\ 0 & 0 & \dots & 0 & 1 & | & 1 \end{pmatrix}$$

Back substitution:

$$x_{n-1} - x_n = -1 \Rightarrow x_{n-1} = x_n - 1 = 0$$

$$x_{n-2} - x_{n-1} - x_n = -1 \Rightarrow x_{n-2} = x_{n-1} + x_n - 1 = 2x_{n-1} = 0$$

$$x_{n-3} - x_{n-2} - x_{n-1} - x_n = -1 \Rightarrow x_{n-3} = x_{n-2} + x_{n-1} + x_n - 1 = 2x_{n-2} = 0$$

..

 $x_n = 1$

Example 4.2

Back substitution

$$x_{n-1} - x_n = -1 \Rightarrow x_{n-1} = x_n - 1 = 0$$

$$x_{n-2} - x_{n-1} - x_n = -1 \Rightarrow x_{n-2} = x_{n-1} + x_n - 1 = 2x_{n-1} = 0$$

$$x_{n-3} - x_{n-2} - x_{n-1} - x_n = -1 \Rightarrow x_{n-3} = x_{n-2} + x_{n-1} + x_n - 1 = 2x_{n-2} = 0$$

$$x_{n-3} - x_{n-2} - x_{n-1} - x_n = -1 \Rightarrow x_{n-3} = x_{n-2} + x_{n-1} + x_n - 1 = 2x_{n-2} = 0$$
...

$$x_{n-k} - x_{n-k+1} - \sum_{i=n-k+2}^{n} x_i = -1 \Rightarrow$$

$$\Rightarrow x_{n-k} = x_{n-k+1} + \sum_{i=n-k+2}^{n} x_i - 1 = 2x_{n-k+1} = 0$$

$$x_{n-k-1} - x_{n-k} - \sum_{i=n-k+1} x_i = -1 \Rightarrow$$

 $\Rightarrow x_{n-k-1} = x_{n-k} + \sum_{i=n-k+1}^{n} x_i - 1 = 2x_{n-k} = 0$

 $x_n = 1, x_{n-1} = 0, x_{n-2} = 0, ..., x_1 = 0$ Ramaz Botchorishvili (KIU)

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Example 4.3

What if we have small error in data, for example $b_n=1+\epsilon$

$$\begin{pmatrix} 1 & -1 & \dots & \dots & -1 \\ 0 & 1 & \dots & \dots & -1 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 1 & -1 \\ 0 & 0 & \dots & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_{n-1} \\ x_n \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \\ \dots \\ -1 \\ 1 + \epsilon \end{pmatrix},$$

$$\begin{pmatrix} 1 & -1 & \dots & \dots & -1 & | & -1 \\ 0 & 1 & \dots & \dots & -1 & | & -1 \\ \dots & \dots & \dots & \dots & | & \dots \\ 0 & 0 & \dots & 1 & -1 & | & -1 \\ 0 & 0 & \dots & 0 & 1 & | & 1+\epsilon \end{pmatrix}$$

Example 4.4

Suppose right hand side contains some error ϵ and $b_n=1+\epsilon$

$$x_{n} = 1 + \epsilon$$

$$x_{n-1} - x_{n} = -1 \Rightarrow x_{n-1} = x_{n} - 1 = 1 + \epsilon - 1 = \epsilon$$

$$x_{n-2} - x_{n-1} - x_{n} = -1 \Rightarrow x_{n-2} = x_{n-1} + x_{n} - 1 = 2x_{n-1} = 2\epsilon$$

$$x_{n-3} - x_{n-2} - x_{n-1} - x_{n} = -1 \Rightarrow x_{n-3} = x_{n-2} + x_{n-1} + x_{n} - 1 = 2x_{n-2}$$

$$= 2^{2} \epsilon$$

$$x_{n-k} = x_{n-k+1} + \sum_{i=n-k+2}^{n} x_i - 1 = 2x_{n-k+1} = 2^{k-1}\epsilon$$

$$x_{n-k-1} - x_{n-k} - \sum_{i=n-k+1}^{n} x_i = -1 \Rightarrow$$

$$x_{n-k-1} = x_{n-k} + \sum_{i=n-k+1}^{n} x_i - 1 = 2x_{n-k} = 2^k \epsilon$$

 $x_n = 1 + \epsilon, x_{n-1} = \epsilon, x_{n-2} = 2^1 \epsilon, x_{n-3} = 2^2 \epsilon, ..., x_1 = 2^{n-1} \epsilon$

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Example 4.5

Suppose right hand side contains some error ϵ and $b_n=1+\epsilon$ $x_n=1+\epsilon, x_{n-1}=\epsilon, x_{n-2}=2^1\epsilon, x_{n-3}=2^2\epsilon, ..., x_1=2^{n-1}\epsilon$

Example 4.5

Suppose right hand side contains some error ϵ and $b_n = 1 + \epsilon$ $x_n = 1 + \epsilon, x_{n-1} = \epsilon, x_{n-2} = 2^1 \epsilon, x_{n-3} = 2^2 \epsilon, ..., x_1 = 2^{n-1} \epsilon$



Figure: Error amplification factors 2^{n-1}

$$\begin{pmatrix} 1 & -1 & \dots & \dots & -1 \\ 0 & 1 & \dots & \dots & -1 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 1 & -1 \\ 0 & 0 & \dots & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_{n-1} \\ x_n \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \\ \dots \\ -1 \\ 1 \end{pmatrix}, x = \begin{pmatrix} 0 \\ 0 \\ \dots \\ 0 \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & -1 & \dots & \dots & -1 \\ 0 & 1 & \dots & \dots & -1 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 1 & -1 \\ 0 & 0 & \dots & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_{n-1} \\ x_n \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \\ \dots \\ -1 \\ 1 \end{pmatrix}, x = \begin{pmatrix} 0 \\ 0 \\ \dots \\ 0 \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & -1 & \dots & -1 \\ 0 & 1 & \dots & -1 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 1 & -1 \\ 0 & 0 & \dots & 0 & 1 \end{pmatrix} \begin{pmatrix} \tilde{x}_1 \\ \tilde{x}_2 \\ \dots \\ \tilde{x}_{n-1} \\ \tilde{x}_n \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \\ \dots \\ -1 \\ 1 + \epsilon \end{pmatrix}, \tilde{x} = \begin{pmatrix} 2^{n-1}\epsilon \\ 2^{n-2}\epsilon \\ \dots \\ \epsilon \\ 1 + \epsilon \end{pmatrix}$$

Triangular systems of linear equations 8, errors

- ► Can we compare two vectors?
- ▶ How do we compute errors in the solution of linear systems?

Triangular systems of linear equations 8, errors

- ► Can we compare two vectors?
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Definition 4.6

Let $x \in \mathbb{R}$ and let $\tilde{x} \in \mathbb{R}$ denote an approximation of it.

- ▶ Absolute error: $|x \tilde{x}|$
- ► Relative error: $\frac{|x-\tilde{x}|}{|x|}$

Triangular systems of linear equations 8, errors

- Can we compare two vectors?
- ▶ How do we compute errors in the solution of linear systems?

Definition 4.6

Let $x \in \mathbb{R}$ and let $\tilde{x} \in \mathbb{R}$ denote an approximation of it.

- ▶ Absolute error: $|x \tilde{x}|$
- ► Relative error: $\frac{|x-\tilde{x}|}{|x|}$

Definition 4.7

Let $x \in \mathbb{R}^n$ and let $\tilde{x} \in \mathbb{R}^n$ denote an approximation of it.

- ▶ Absolute error: $||x \tilde{x}||$
- ► Relative error: $\frac{\|x \tilde{x}\|}{\|x\|}$

Triangular systems of linear equations 9, errors

- ► Can we compare two vectors?
- ▶ How do we compute errors in the solution of linear systems?

$$x = \begin{pmatrix} 0 \\ 0 \\ \dots \\ 0 \\ 1 \end{pmatrix}, \tilde{x} = \begin{pmatrix} 2^{n-1}\epsilon \\ 2^{n-2}\epsilon \\ \dots \\ \epsilon \\ 1+\epsilon \end{pmatrix}, \quad \tilde{x} - x = \begin{pmatrix} 2^{n-1}\epsilon \\ 2^{n-2}\epsilon \\ \dots \\ \epsilon \\ \epsilon \end{pmatrix}, ||\tilde{x} - x|| = 2^{n-1}|\epsilon|$$

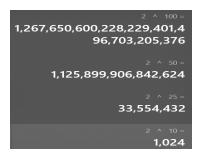


Figure: Error amplification factors 2^{n-1}

Triangular systems of linear equations 10, ill conditioned matrix

Example 4.8

- lackbox Two systems of linear equations with the same matrix A, Ax=b and $A ilde{x}= ilde{b}$
- ightharpoonup b and $ilde{b}$ are two different inputs for the same problem
- ▶ Well conditioned problem: if b and \tilde{b} are close then x and \tilde{x} are close.
- ▶ III conditioned problem: even if b and \tilde{b} are close then x and \tilde{x} differ from each other drastically.

Triangular systems of linear equations 10, ill conditioned matrix

Example 4.8

- \blacktriangleright Two systems of linear equations with the same matrix A, Ax = b and $A\tilde{x} = \tilde{b}$
- ightharpoonup b and \tilde{b} are two different inputs for the same problem
- ▶ Well conditioned problem: if b and \tilde{b} are close then x and \tilde{x} are close.
- ▶ III conditioned problem: even if b and \tilde{b} are close then x and \tilde{x} differ from each other drastically.

Theorem 4.9

The problem
$$Ax = b$$
 is ill posed, if $A = \begin{pmatrix} 1 & -1 & \dots & -1 \\ 0 & 1 & \dots & -1 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 1 & -1 \\ 0 & 0 & \dots & 0 & 1 \end{pmatrix}$

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Theorem 4.10

The problem
$$Ax = b$$
 is ill posed, if $A = \begin{pmatrix} 1 & -1 & \dots & -1 \\ 0 & 1 & \dots & -1 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 1 & -1 \\ 0 & 0 & \dots & 0 & 1 \end{pmatrix}$

Theorem 4.10

The problem
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Which matrix properties are important for conditioning?

▶ Determinant of a matrix?

Theorem 4.10

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$$Ax = b$$
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Which matrix properties are important for conditioning?

- ▶ Determinant of a matrix?
- ► Eigenvalue of a matrix?

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$$Ax = b$$
 is ill posed, if $A = \begin{pmatrix} 1 & -1 & \dots & -1 \\ 0 & 1 & \dots & -1 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 1 & -1 \\ 0 & 0 & \dots & 0 & 1 \end{pmatrix}$

Which matrix properties are important for conditioning?

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- ▶ Big or small entries of a matrix?

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Which matrix properties are important for conditioning?

- Determinant of a matrix?
- Eigenvalue of a matrix?
- ▶ Big or small entries of a matrix?
- ▶ Big or small norm of a matrix?

Definition 4.11

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

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Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.12

▶ Ill-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 4.001 & 2.002 \\ 1 & 2.002 & 2.004 \end{pmatrix}, b = \begin{pmatrix} 1 \\ 8.0021 \\ 5.006 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

Definition 4.11

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.12

▶ Ill-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 4.001 & 2.002 \\ 1 & 2.002 & 2.004 \end{pmatrix}, b = \begin{pmatrix} 1 \\ 8.0021 \\ 5.006 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

$$\tilde{b} = \begin{pmatrix} 1 \\ 8.0020 \\ 5.0061 \end{pmatrix}, \tilde{x} = \begin{pmatrix} 3.0850 \\ -0.0436 \\ 1.0022 \end{pmatrix}$$

Definition 4.11

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Example 4.12

▶ Ill-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 4.001 & 2.002 \\ 1 & 2.002 & 2.004 \end{pmatrix}, b = \begin{pmatrix} 1 \\ 8.0021 \\ 5.006 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

$$\tilde{b} = \begin{pmatrix} 1 \\ 8.0020 \\ 5.0061 \end{pmatrix}, \tilde{x} = \begin{pmatrix} 3.0850 \\ -0.0436 \\ 1.0022 \end{pmatrix}$$

Small relative perturbation: $\frac{\|b-\tilde{b}\|}{\|b\|} = 1.3975 \cdot 10^{-5}$

Large relative error: $\frac{\|x - \tilde{x}\|}{\|x\|} = 1.3461$

Definition 4.13

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Definition 4.13

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.14

▶ III-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 4.001 & 2.002 \\ 1 & 2.002 & 2.004 \end{pmatrix}, b = \begin{pmatrix} 1 \\ 8.0021 \\ 5.006 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

Definition 4.13

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.14

▶ Ill-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 4.001 & 2.002 \\ 1 & 2.002 & 2.004 \\ \end{pmatrix}, b = \begin{pmatrix} 1 \\ 8.0021 \\ 5.006 \\ \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \\ 1 \\ \end{pmatrix}$$

$$A = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4.001 & 2.002 \\ [1. & 2.002 & 2.004] \\ [1. & 2.002 & 2.004] \\ [1. & 2.002 & 2.004] \\ [2. & 4.001 & 31326.00296885506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326.0029688506 \\ [3. & 31326$$

Definition 4.13

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.14

▶ Ill-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 4.001 & 2.002 \\ 1 & 2.002 & 2.004 \end{pmatrix}, b = \begin{pmatrix} 1 \\ 8.0021 \\ 5.006 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

$$A = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4.001 & 2.002 \\ [1. & 2.002 & 2.004] \\ [1. & 2.002 & 2.004] \end{bmatrix}$$

$$cond2(A) = 31062.1661044696$$

$$condFro(A) = 31326.00296885506$$

$$condInf(A) = 48170.05699998381$$

$$cond1(A) = 48170.05699998381$$

Small relative perturbation: $\frac{\|b-\tilde{b}\|}{\|b\|}=1.3975\cdot 10^{-5}$

Large relative error: $\frac{\|x-\tilde{x}\|}{\|x\|} = 1.3461$

Definition 4.15

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.16

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Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.16

▶ Well-conditioned system of linear equations Ax = b

Definition 4.15

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.16

▶ Well-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}, b = \begin{pmatrix} 3 \\ 7 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

Definition 4.15

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.16

▶ Well-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}, b = \begin{pmatrix} 3 \\ 7 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\tilde{b} = \begin{pmatrix} 3.0001 \\ 7.0001 \end{pmatrix}, \tilde{x} = \begin{pmatrix} 0.9999 \\ 1.0001 \end{pmatrix}$$

Definition 4.15

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.16

ightharpoonup Well-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}, b = \begin{pmatrix} 3 \\ 7 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\tilde{b} = \begin{pmatrix} 3.0001 \\ 7.0001 \end{pmatrix}, \tilde{x} = \begin{pmatrix} 0.9999 \\ 1.0001 \end{pmatrix}$$

Small relative perturbation: $\frac{\|b-\tilde{b}\|}{\|b\|} = 1.875 \cdot 10^{-5}$

Definition 4.15

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.16

▶ Well-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}, b = \begin{pmatrix} 3 \\ 7 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\tilde{b} = \begin{pmatrix} 3.0001 \\ 7.0001 \end{pmatrix}, \tilde{x} = \begin{pmatrix} 0.9999 \\ 1.0001 \end{pmatrix}$$

Small relative perturbation: $\frac{\|b-\tilde{b}\|}{\|b\|} = 1.875 \cdot 10^{-5}$

Small relative error: $\frac{\|x-\tilde{x}\|}{\|x\|} = 10^{-5}$

Definition 4.15

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Example 4.16

▶ Well-conditioned system of linear equations Ax = b

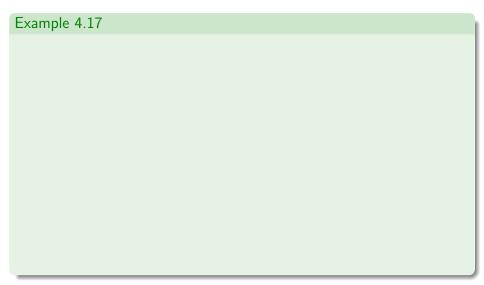
$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}, b = \begin{pmatrix} 3 \\ 7 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\tilde{b} = \begin{pmatrix} 3.0001 \\ 7.0001 \end{pmatrix}, \tilde{x} = \begin{pmatrix} 0.9999 \\ 1.0001 \end{pmatrix}$$

Small relative perturbation: $\frac{\|b-\tilde{b}\|}{\|b\|} = 1.875 \cdot 10^{-5}$

Small relative error: $\frac{\|x-\tilde{x}\|}{\|x\|} = 10^{-5}$

small condition number: cond(A) = 14.9930



Example 4.17

▶ Well-conditioned system of linear equations Ax = b

Example 4.17

ightharpoonup Well-conditioned system of linear equations Ax = b

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}, b = \begin{pmatrix} 3 \\ 7 \end{pmatrix}, x = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

Example 4.17

▶ Well-conditioned system of linear equations Ax = b

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▶ Well-conditioned system of linear equations Ax = b

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Small relative perturbation: $\frac{\|b-\tilde{b}\|}{\|b\|}=1.875\cdot 10^{-5}$ Small relative error: $\frac{\|x-\tilde{x}\|}{\|x\|}=10^{-5}$

small condition number:

```
A= [[1 2]

[3 4]]

cond2(A)= 14.933034373659265

condFro(A)= 14.999999999999998

condInf(A)= 21.0

cond1(A)= 20.999999999999999
```

Definition 4.18

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Condition number of a matrix may be different in different norms

Definition 4.18

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Condition number of a matrix may be different in different norms

Theorem 4.19

1. $cond(A) \ge 1$ for any induced norm

Definition 4.18

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Condition number of a matrix may be different in different norms

Theorem 4.19

1. $cond(A) \ge 1$ for any induced norm (Solution: $A \cdot A^{-1} = I$)

Definition 4.18

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Condition number of a matrix may be different in different norms

- 1. $cond(A) \ge 1$ for any induced norm (Solution: $A \cdot A^{-1} = I$)
- 2. $cond(\alpha A) = cond(A)$ for any norm

Definition 4.18

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Condition number of a matrix may be different in different norms

- 1. $cond(A) \ge 1$ for any induced norm (Solution: $A \cdot A^{-1} = I$)
- 2. $cond(\alpha A) = cond(A)$ for any norm (Solution: $\alpha A \cdot \alpha^{-1} A^{-1} = I$)

Definition 4.18

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Condition number of a matrix may be different in different norms

- 1. $cond(A) \ge 1$ for any induced norm (Solution: $A \cdot A^{-1} = I$)
- 2. $cond(\alpha A) = cond(A)$ for any norm (Solution: $\alpha A \cdot \alpha^{-1} A^{-1} = I$)
- 3. $cond(AB) \le cond(A)cond(B)$ for any sub-multiplicative norm

Definition 4.18

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Condition number of a matrix may be different in different norms

- 1. $cond(A) \ge 1$ for any induced norm (Solution: $A \cdot A^{-1} = I$)
- 2. $cond(\alpha A) = cond(A)$ for any norm (Solution: $\alpha A \cdot \alpha^{-1} A^{-1} = I$)
- 3. $cond(AB) \le cond(A)cond(B)$ for any sub-multiplicative norm
- 4. $cond_1(A) = cond_{\infty}(A^T)$

Definition 4.18

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Condition number of a matrix may be different in different norms

- 1. $cond(A) \ge 1$ for any induced norm (Solution: $A \cdot A^{-1} = I$)
- 2. $cond(\alpha A) = cond(A)$ for any norm (Solution: $\alpha A \cdot \alpha^{-1} A^{-1} = I$)
- 3. $cond(AB) \le cond(A)cond(B)$ for any sub-multiplicative norm
- 4. $cond_1(A) = cond_{\infty}(A^T)$
- 5. $cond(A) = cond(A^{-1})$

Definition 4.20

Condition number of a matrix: $cond(A) = ||A|| ||A^{-1}||$

Condition number computed in one norm and condition number computed in another may be different

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 iff $A^T A = \alpha I$, $\alpha \neq 0$

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condition number characterize ill- and well-conditioned problems

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Example 4.22

lacktriangle Two systems of linear equations with the same matrix A, Ax=b and $A ilde{x}= ilde{b}$

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$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \circ \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \end{bmatrix} = \begin{bmatrix} a_{11} b_{11} & a_{12} b_{12} & a_{13} b_{13} \\ a_{21} b_{21} & a_{22} b_{22} & a_{23} b_{23} \\ a_{31} b_{31} & a_{32} b_{32} & a_{33} b_{33} \end{bmatrix}$$

Figure: Jacques Hadamard, Hadamard product, source Wikipedia

- ▶ J.Hadamard concept of well posed problem:
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Theorem 4.23

(Right perturbation theorem) Suppose

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(Right perturbation theorem) Suppose

- ► A is invertible
- ightharpoonup Ax = b
- \triangleright δb is perturbation of b
- δx is pertubation caused by δb

Then the following holds true:

$$\frac{1}{\|A\|\|A^{-1}\|} \frac{\|\delta b\|}{\|b\|} \le \frac{\|\delta x\|}{\|x\|} \le \|A\| \|A^{-1}\| \frac{\|\delta b\|}{\|b\|}$$

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$$\frac{1}{cond(A)} \frac{\|\delta b\|}{\|b\|} \leq \frac{\|\delta x\|}{\|x\|} \leq cond(A) \frac{\|\delta b\|}{\|b\|}$$

Proof.

Proof.

$$ightharpoonup Ax = b$$

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- ightharpoonup Ax = b
- $A(x + \delta x) = b + \delta b$

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- $ightharpoonup A\delta x = \delta b$
- $\delta x = A^{-1} \delta b$
- $\|\delta x\| = \|A^{-1}\delta b\| \le \|A^{-1}\| \|\delta b\|$

Proof.

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- $ightharpoonup A(x + \delta x) = b + \delta b$
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- $||b|| = ||Ax|| \le ||A|| ||x||$

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Proof.

- ightharpoonup Ax = b
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- $ightharpoonup A\delta x = \delta b$
- $\|\delta x\| = \|A^{-1}\delta b\| \le \|A^{-1}\| \|\delta b\|$
- $\|b\| = \|Ax\| \le \|A\| \|x\|$
- $\|\delta x\|\|b\| \le \|A^{-1}\|\|\delta b\|\|A\|\|x\|$
- $\qquad \qquad \blacksquare \frac{\|\delta x\|}{\|x\|} \le \|A\| \|A^{-1}\| \frac{\|\delta b\|}{\|b\|} = cond(A) \frac{\|\delta b\|}{\|b\|}$

Proof.

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$$\|\delta x\| = \|A^{-1}\delta b\| \le \|A^{-1}\| \|\delta b\|$$

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$$\|\delta x\| \|b\| \le \|A^{-1}\| \|\delta b\| \|A\| \|x\|$$

$$\frac{1}{\|A\| \|A^{-1}\|} \frac{\|\delta b\|}{\|b\|} \le \frac{\|\delta x\|}{\|x\|} \le \|A\| \|A^{-1}\| \frac{\|\delta b\|}{\|b\|}$$



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- $\triangleright x = A^{-1}b$
- $\|x\| = \|A^{-1}b\| \le \|A^{-1}\|\|b\|$

Proof.

- ightharpoonup Ax = b
- $ightharpoonup A(x + \delta x) = b + \delta b$
- $ightharpoonup A\delta x = \delta b$
- ► $x = A^{-1}b$
- $||x|| = ||A^{-1}b|| \le ||A^{-1}|| ||b||$
- $\|\delta b\| \|x\| \le \|A\| \|\delta x\| \|A^{-1}\| \|b\|$

Proof.

- ightharpoonup Ax = b
- $ightharpoonup A(x + \delta x) = b + \delta b$
- $ightharpoonup A\delta x = \delta b$
- $x = A^{-1}b$
- $\|x\| = \|A^{-1}b\| \le \|A^{-1}\|\|b\|$
- $\|\delta b\| \|x\| \le \|A\| \|\delta x\| \|A^{-1}\| \|b\|$
- $\qquad \qquad \blacksquare \frac{\|\delta b\|}{\|b\|} \le \|A\| \|A^{-1}\| \frac{\|\delta x\|}{\|x\|} = cond(A) \frac{\|\delta x\|}{\|x\|}$

Proof.

(Right perturbation theorem, part 2)

- ightharpoonup Ax = b
- $ightharpoonup A(x + \delta x) = b + \delta b$
- $ightharpoonup A\delta x = \delta b$
- $x = A^{-1}b$
- $\|\delta b\| = \|A\delta x\| \le \|A\| \|\delta x\|$
- $\|x\| = \|A^{-1}b\| \le \|A^{-1}\|\|b\|$
- $\|\delta b\|\|x\| \le \|A\|\|\delta x\|\|A^{-1}\|\|b\|$
- $| \frac{\|\delta b\|}{\|b\|} \le \|A\| \|A^{-1}\| \frac{\|\delta x\|}{\|x\|} = cond(A) \frac{\|\delta x\|}{\|x\|}$

$$\frac{1}{\|A\| \|A^{-1}\|} \frac{\|\delta b\|}{\|b\|} \le \frac{\|\delta x\|}{\|x\|} \le \|A\| \|A^{-1}\| \frac{\|\delta b\|}{\|b\|}$$



Right perturbation theorem, questions, discussion

$$\frac{1}{\|A\|\|A^{-1}\|}\frac{\|\delta b\|}{\|b\|} \le \frac{\|\delta x\|}{\|x\|} \le \|A\|\|A^{-1}\|\frac{\|\delta b\|}{\|b\|}$$

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$$\frac{1}{\|A\|\|A^{-1}\|}\frac{\|\delta b\|}{\|b\|} \leq \frac{\|\delta x\|}{\|x\|} \leq \|A\|\|A^{-1}\|\frac{\|\delta b\|}{\|b\|}$$

► Left inequality

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- Left inequality
- Right inequality

Right perturbation theorem, questions, discussion

$$\frac{1}{\|A\|\|A^{-1}\|}\frac{\|\delta b\|}{\|b\|} \leq \frac{\|\delta x\|}{\|x\|} \leq \|A\|\|A^{-1}\|\frac{\|\delta b\|}{\|b\|}$$

- ► Left inequality
- Right inequality
- can lower and upper bounds reached?

Theorem 4.24

(Left perturbation theorem) Suppose

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- $\|\delta A\| < 1/\|A^{-1}\|$

Theorem 4.24

(Left perturbation theorem) Suppose

- ► A is invertible
- ightharpoonup Ax = b
- \triangleright δA is perturbation of A
- \blacktriangleright δx is pertubation caused by δA
- $ightharpoonup \|\delta A\| < 1/\|A^{-1}\|$

Then the following holds true:

$$\frac{\|\delta x\|}{\|x\|} \le \frac{\|A^{-1}\| \|\delta A\|}{1 - \|A^{-1}\| \|\delta A\|}$$

Theorem 4.24

(Left perturbation theorem)
Suppose

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- \triangleright Ax = b
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Then the following holds true:

$$\frac{\|\delta x\|}{\|x\|} \le \frac{\|A^{-1}\| \|\delta A\|}{1 - \|A^{-1}\| \|\delta A\|}$$

$$\frac{\|\delta x\|}{\|x\|} \le \frac{\operatorname{cond}(A) \frac{\|\delta A\|}{\|A\|}}{1 - \operatorname{cond}(A) \frac{\|\delta A\|}{\|\delta A\|}}$$

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- $Ax + A\delta x + \delta Ax + \delta A\delta x = b$

Proof.

- ightharpoonup Ax = b
- $(A + \delta A)(x + \delta x) = b$
- $Ax + A\delta x + \delta Ax + \delta A\delta x = b$
- $A\delta x + \delta Ax + \delta A\delta x = 0$

Proof.

- ightharpoonup Ax = b
- $(A + \delta A)(x + \delta x) = b$
- $Ax + A\delta x + \delta Ax + \delta A\delta x = b$
- $A\delta x + \delta Ax + \delta A\delta x = 0$
- $A\delta x = -\delta A(x + \delta x)$

Proof.

- ightharpoonup Ax = b
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- $(A + \delta A)(x + \delta x) = b$
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- $A\delta x + \delta A x + \delta A \delta x = 0$
- $A\delta x = -\delta A(x + \delta x)$
- $\|\delta x\| \le \|A^{-1}\| \|\delta A\| (\|x\| + \|\delta x\|)$

Proof.

- ightharpoonup Ax = b
- $(A + \delta A)(x + \delta x) = b$
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- $\|\delta x\| \le \|A^{-1}\| \|\delta A\| (\|x\| + \|\delta x\|)$
- $(1 \|A^{-1}\| \|\delta A\|) \|\delta x\| \le \|A^{-1}\| \|\delta A\| \|x\|$

Proof.

- ightharpoonup Ax = b
- $(A + \delta A)(x + \delta x) = b$
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- $A\delta x = -\delta A(x + \delta x)$
- $\|\delta x\| \le \|A^{-1}\| \|\delta A\| (\|x\| + \|\delta x\|)$
- $| (1 ||A^{-1}|||\delta A||) ||\delta x|| \le ||A^{-1}|||\delta A|||x||$

$$\frac{\|\delta x\|}{\|x\|} \le \frac{\|A^{-1}\| \|\delta A\|}{1 - \|A^{-1}\| \|\delta A\|} \equiv \frac{\|\delta x\|}{\|x\|} \le \frac{cond(A) \frac{\|\delta A\|}{\|A\|}}{1 - cond(A) \frac{\|\delta A\|}{\|A\|}}$$

Q & A