### **Sensitivity and Conditioning**

CS 205A:

Mathematical Methods for Robotics, Vision, and Graphics

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### Questions

Gaussian elimination works in theory, but what about floating point precision?

How much can we trust  $\vec{x}_0$  if  $0 < ||A\vec{x}_0 - \vec{b}|| \ll 1$ ?

### Recall: Backward Error

### **Backward Error**

The amount a problem statement would have to change to realize a given approximation of its solution

**Example 1:**  $\sqrt{x}$ 

**Example 2:**  $A\vec{x} = \vec{b}$ 

### **Perturbation Analysis**

How does 
$$\vec{x}$$
 change if we solve  $(A + \delta A)\vec{x} = \vec{b} + \delta \vec{b}$ ?

### Two viewpoints:

- ▶ Thanks to floating point precision, A and  $\vec{b}$  are approximate
- If  $\vec{x}_0$  isn't the exact solution, what is the backward error?



### What is "Small?"

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#### Vector norm

A function  $\|\cdot\|:\mathbb{R}^n\to[0,\infty)$  satisfying:

- **1.**  $\|\vec{x}\| = 0$  iff  $\vec{x} = 0$
- **2.**  $||c\vec{x}|| = |c|||\vec{x}|| \ \forall c \in R, \vec{x} \in \mathbb{R}^n$
- **3.**  $\|\vec{x} + \vec{y}\| < \|\vec{x}\| + \|\vec{y}\| \ \forall \vec{x}, \vec{y} \in \mathbb{R}^n$

### Our Favorite Norm

$$\|\vec{x}\|_2 \equiv \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$

### p-Norms

For 
$$p \geq 1$$
,

$$\|\vec{x}\|_p \equiv (|x_1|^p + |x_2|^p + \dots + |x_n|^p)^{1/p}$$

Taxicab norm:  $\|\vec{x}\|_1$ 



$$\|\vec{x}\|_{\infty} \equiv \max(|x_1|, |x_2|, \dots, |x_n|)$$

<unit\_circles>

### How are Norms the Same?

### **Equivalent norms**

Two norms  $\|\cdot\|$  and  $\|\cdot\|'$  are equivalent if there exist constants  $c_{low}$  and  $c_{high}$  such that

$$|c_{low}||\vec{x}|| \le ||\vec{x}||' \le c_{high}||\vec{x}||$$
 for all  $\vec{x} \in \mathbb{R}^n$ .

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### **Theorem**

All norms on  $\mathbb{R}^n$  are equivalent.

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### **Theorem**

All norms on  $\mathbb{R}^n$  are equivalent.

(10000, 1000, 1000) vs. (10000, 0, 0)?

# Matrix Norms: "Unrolled" Construction

$$A \in \mathbb{R}^{m \times n} \leftrightarrow \mathtt{a(:)} \in \mathbb{R}^{mn}$$

$$||A||_{\text{Fro}} \equiv \sqrt{\sum_{ij} a_{ij}^2}$$

## Matrix Norms: "Induced" Construction

$$||A|| \equiv \max\{||A\vec{x}|| : ||\vec{x}|| = 1\}$$

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What is the norm induced by  $\|\cdot\|_2$ ?

### **Other Induced Norms**

$$||A||_1 \equiv \max_j \sum_i |a_{ij}|$$

$$||A||_{\infty} \equiv \max_i \sum_i |a_{ij}|$$

### Question

### Are all matrix norms equivalent?

### Recall: Condition Number

### Condition number

Ratio of forward to backward error

### **Root-finding example:**

$$\frac{1}{f'(x^*)}$$

### **Model Problem**

$$(A + \varepsilon \cdot \delta A)\vec{x}(\varepsilon) = \vec{b} + \varepsilon \cdot \delta \vec{b}$$

### Simplification (on the board!)

$$\frac{d\vec{x}}{d\varepsilon}\Big|_{\varepsilon=0} = A^{-1}(\delta\vec{b} - \delta A \cdot \vec{x}(0))$$

$$\frac{\|\vec{x}(\varepsilon) - \vec{x}(0)\|}{\|\vec{x}(0)\|} \le |\varepsilon| \|A^{-1}\| \|A\| \left( \frac{\|\delta\vec{b}\|}{\|\vec{b}\|} + \frac{\|\delta A\|}{\|A\|} \right) + O(\varepsilon^2)$$

### Condition Number

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The condition number of  $A \in \mathbb{R}^{n \times n}$  for a given matrix norm  $\|\cdot\|$  is  $\operatorname{cond} A \equiv \kappa \equiv \|A^{-1}\| \|A\|$ .

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Relative change: 
$$D \equiv \frac{\delta \vec{b}}{\|\vec{b}\|} + \frac{\|\delta A\|}{\|A\|}$$
 
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Invariant to scaling (unlike determinant!); equals one for the identity.

### **Condition Number of Induced Norm**

cond 
$$A = \left(\max_{\vec{x} \neq \vec{0}} \frac{\|A\vec{x}\|}{\|\vec{x}\|}\right) \left(\min_{\vec{y} \neq \vec{0}} \frac{\|A\vec{y}\|}{\|\vec{y}\|}\right)^{-1}$$

### **Chicken** ← **Egg**

$$\operatorname{cond} A \equiv \|A\| \overline{\|A^{-1}\|}$$

Computing  $||A^{-1}||$  typically requires solving  $A\vec{x} = \vec{b}$ , but how do we know the reliability of  $\vec{x}$ ?

### To Avoid...

What is the condition number of computing the condition number of A?

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### Instead

# Bound the condition number.

- Below: Problem is at least this hard
- Above: Problem is at most this hard

### **Potential for Approximation**

$$||A^{-1}\vec{x}|| \le ||A^{-1}|| ||\vec{x}||$$

$$\downarrow \qquad \qquad \qquad \downarrow$$

$$\text{cond } A = ||A|| ||A^{-1}|| \ge \frac{||A|| ||A^{-1}\vec{x}||}{||\vec{x}||}$$