# 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

$$
0<\left\|A \vec{x}_{0}-\vec{b}\right\| \ll 1 ?
$$

## 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?"

What does it mean for a statement to hold for small $\delta \vec{x}$ ?

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## Vector norm

A function $\|\cdot\|: \mathbb{R}^{n} \rightarrow[0, \infty)$ 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. $\|\vec{x}+\vec{y}\| \leq\|\vec{x}\|+\|\vec{y}\| \forall \vec{x}, \vec{y} \in \mathbb{R}^{n}$

## Our Favorite Norm

$$
\|\vec{x}\|_{2} \equiv \sqrt{x_{1}^{2}+x_{2}^{2}+\cdots+x_{n}^{2}}
$$

## p-Norms

$$
\begin{gathered}
\text { For } p \geq 1 \\
\|\vec{x}\|_{p} \equiv\left(\left|x_{1}\right|^{p}+\left|x_{2}\right|^{p}+\cdots+\left|x_{n}\right|^{p}\right)^{1 / p} \\
\text { Taxicab norm: }\|\vec{x}\|_{1}
\end{gathered}
$$

## $\infty$ Norm

$$
\|\vec{x}\|_{\infty} \equiv \max \left(\left|x_{1}\right|,\left|x_{2}\right|, \ldots,\left|x_{n}\right|\right)
$$

## How are Norms Different?

## <unit_circles>

## How are Norms the Same?

## Equivalent norms

Two norms $\|\cdot\|$ and $\|\cdot\|^{\prime}$ are equivalent if there exist constants $c_{\text {low }}$ and $c_{\text {high }}$ such that $c_{\text {low }}\|\vec{x}\| \leq\|\vec{x}\|^{\prime} \leq c_{h i g h}\|\vec{x}\|$ 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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All norms on $\mathbb{R}^{n}$ are equivalent.

$$
(10000,1000,1000) \text { vs. }(10000,0,0) ?
$$

## Matrix Norms:

 "Unrolled" Construction$$
A \in \mathbb{R}^{m \times n} \leftrightarrow \mathrm{a}(:) \in \mathbb{R}^{m n}
$$



## 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

$$
\begin{aligned}
\|A\|_{1} & \equiv \max _{j} \sum_{i}\left|a_{i j}\right| \\
\|A\|_{\infty} & \equiv \max _{i} \sum_{j}\left|a_{i j}\right|
\end{aligned}
$$

## Question

## Are all matrix norms equivalent?

## Recall: Condition Number

## Condition number <br> Ratio of forward to backward error

## Root-finding example:

$$
\frac{1}{f^{\prime}\left(x^{*}\right)}
$$

## Model Problem

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

## Simplification (on the board!)

$$
\begin{gathered}
\left.\frac{d \vec{x}}{d \varepsilon}\right|_{\varepsilon=0}=A^{-1}(\delta \vec{b}-\delta A \cdot \vec{x}(0)) \\
\frac{\|\vec{x}(\varepsilon)-\vec{x}(0)\|}{\|\vec{x}(0)\|} \leq|\varepsilon|\left\|A^{-1}\right\|\|A\|\left(\frac{\|\delta \vec{b}\|}{\|\vec{b}\|}+\frac{\|\delta A\|}{\|A\|}\right)+O\left(\varepsilon^{2}\right)
\end{gathered}
$$

## Condition Number

## Condition number <br> The condition number of $A \in \mathbb{R}^{n \times n}$ for a given matrix norm $\|\cdot\|$ is cond $A \equiv \kappa \equiv\left\|A^{-1}\right\|\|A\|$.

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$$

Invariant to scaling (unlike determinant!); equals one for the identity.

## Condition Number of Induced Norm



## Chicken $\Longleftrightarrow$ Egg

## cond $A \equiv\|A\|\left\|A^{-1}\right\|$

## Computing $\left|\left|A^{-1}\right|\right|$ 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

$$
\begin{aligned}
\left\|A^{-1} \vec{x}\right\| & \leq\left\|A^{-1}\right\|\|\vec{x}\| \\
& \Downarrow \\
\operatorname{cond} A=\|A\|\left\|A^{-1}\right\| & \geq \frac{\|A\|\left\|A^{-1} \vec{x}\right\|}{\|\vec{x}\|}
\end{aligned}
$$

