

CS 450: Numerical Analysis

Lecture 3

Chapter 2 – Linear Systems

Matrix Norms and Conditioning

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January 26, 2018

Vector Norms

► Properties of vector norms

$$\|\mathbf{x}\| \geq 0$$

$$\|\mathbf{x}\| = 0 \iff \mathbf{x} = \mathbf{0}$$

$$\|\alpha\mathbf{x}\| = |\alpha| \cdot \|\mathbf{x}\|$$

$$\|\mathbf{x} + \mathbf{y}\| \leq \|\mathbf{x}\| + \|\mathbf{y}\| \quad (\textit{triangle inequality}) \textit{ implies continuity}$$

► p -norms

$$\|\mathbf{x}\|_p = \left(\sum_i |x_i|^p \right)^{1/p}$$

Inner-Product Spaces

- ▶ **Properties of inner-product spaces:** Inner products $\langle \mathbf{x}, \mathbf{y} \rangle$ must satisfy

$$\langle \mathbf{x}, \mathbf{x} \rangle \geq 0$$

$$\langle \mathbf{x}, \mathbf{x} \rangle = 0 \iff \mathbf{x} = \mathbf{0}$$

$$\langle \mathbf{x}, \mathbf{y} \rangle = \langle \mathbf{y}, \mathbf{x} \rangle$$

$$\langle \mathbf{x}, \mathbf{y} + \mathbf{z} \rangle = \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{z} \rangle$$

$$\langle \alpha \mathbf{x}, \mathbf{y} \rangle = \alpha \langle \mathbf{x}, \mathbf{y} \rangle$$

- ▶ **Inner-product-based vector norms**

$$\sqrt{\langle \mathbf{x}, \mathbf{x} \rangle}$$

is an norm, due to Cauchy-Schwartz inequality

$$|\langle \mathbf{x}, \mathbf{y} \rangle| \leq \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle \cdot \langle \mathbf{y}, \mathbf{y} \rangle}$$

Inner-products can be expressed as $\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x}^T \mathbf{A} \mathbf{y}$ where \mathbf{A} is symmetric positive definite, yielding norms $\|\mathbf{x}\|_{\mathbf{A}} = \sqrt{\mathbf{x}^T \mathbf{A} \mathbf{x}}$

Matrix Norms

- ▶ **Properties of matrix norms:**

$$\|\mathbf{A}\| \geq 0$$

$$\|\mathbf{A}\| = 0 \Leftrightarrow \mathbf{A} = \mathbf{0}$$

$$\|\alpha\mathbf{A}\| = |\alpha| \cdot \|\mathbf{A}\|$$

$$\|\mathbf{A} + \mathbf{B}\| \leq \|\mathbf{A}\| + \|\mathbf{B}\| \quad (\textit{triangle inequality})$$

- ▶ **Frobenius norm:**

$$\|\mathbf{A}\|_F = \left(\sum_{i,j} a_{ij}^2 \right)^{1/2}$$

- ▶ **Operator/induced/subordinate matrix norms:**

For any vector norm $\|\cdot\|$, the induced matrix norm is

$$\|\mathbf{A}\| = \max_{\mathbf{x} \neq \mathbf{0}} \|\mathbf{Ax}\| / \|\mathbf{x}\| = \max_{\|\mathbf{x}\|=1} \|\mathbf{Ax}\|$$

Induced Matrix Norms

- ▶ **General induced matrix norms:**

$$\|\mathbf{A}\|_{mp} = \max_{\|\mathbf{x}\|_p=1} \|\mathbf{Ax}\|_m$$

*typically $m = p$ so we write $\|\mathbf{A}\|_p$ and almost always we have $p \in \{1, 2, \infty\}$.
(Computing the matrix norm for certain choices of $m \neq p$ is NP-complete.)*

- ▶ **Interpreting induced matrix norms:**

$$\|\mathbf{A}\|_p = \max_{\|\mathbf{x}\|_p=1} \|\mathbf{Ax}\|_p$$

*is the maximum possible p -norm **amplification** due to application of \mathbf{A}*

$$1/\|\mathbf{A}^{-1}\|_p = \min_{\|\mathbf{x}\|_p=1} \|\mathbf{Ax}\|_p$$

*is the maximum possible p -norm **reduction** due to application of \mathbf{A}*

Matrix Condition Number

▶ **Definition:** $\kappa(\mathbf{A}) = \|\mathbf{A}\| \cdot \|\mathbf{A}^{-1}\|$

▶ **Intuitive derivation:**

$$\kappa(\mathbf{A}) = \max_{\text{inputs}} \max_{\text{perturbations in input}} \left| \frac{\text{relative perturbation in output}}{\text{relative perturbation in input}} \right|$$

since a matrix is a linear operator, we can generally decouple its action on the input and the perturbation, so

$$\kappa(\mathbf{A}) = \left| \frac{\overbrace{\max_{\text{perturbations in input}} \text{relative perturbation growth}}^{\|\mathbf{A}\|}}{\underbrace{\max_{\text{inputs}} \text{relative input reduction}}_{1/\|\mathbf{A}^{-1}\|}} \right|$$