

Lecture 4: Introduction

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February 01, 2021

Outline

- 1 Matrix Norms
 - Characterizing the ∞ -norm
 - Characterizing the 2-norm

- 2 Positive definite matrices
 - Definition and example

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∞ -norm

Theorem: Computing the ∞ -norm

- 1 For an $A \in \mathbb{R}^{m \times n}$,

$$\|A\|_1 = \|A'\|_\infty.$$

- 2 Let b'_1, \dots, b'_m be the rows of A . Then

$$\|A\|_\infty = \max_{1 \leq i \leq m} \|b'_i\|_1.$$

∞ -norm

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Comment: The equivalence of the two conditions above follows directly from the characterization of the 1-norm.

Comment: Assertion (2) above can be proved directly, using a similar approach (but with different details) to the proof of the 1-norm case.

∞ -norm

We show how to prove $\|A\|_1 = \|A'\|_\infty$ directly from basic Linear Algebra principles.

Step I: Show that, for any $v \in \mathbb{R}^m$,

$$\|v\|_1 = \max\{(v, w) : \|w\|_\infty = 1\}, \quad \text{and}$$

$$\|v\|_\infty = \max\{(v, w) : \|w\|_1 = 1\}.$$

Step II: Since $\|A\|_1 = \max\{\|Av\|_1 : \|v\|_1 = 1\}$, it follows that

$$\|A\|_1 = \max\{(Av, w) : \|v\|_1 = 1, \|w\|_\infty = 1\}.$$

Step III: Since $\|A'\|_\infty = \max\{\|A'w\|_\infty : \|w\|_\infty = 1\}$, it follows that

$$\|A'\|_\infty = \max\{(A'w, v) : \|v\|_1 = 1, \|w\|_\infty = 1\}.$$

The 2-norm of a matrix

Some basics:

$$\|v\|_2^2 = (v, v).$$

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Whatever A is, $A'A$ is symmetric, and its eigenvalues are non-negative.

$$(BC)' = C'B' \implies (A'A)' = A'A'' = A'A.$$

The 2-norm of a matrix

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$$\|v\|_2^2 = (v, v).$$

Whatever A is, $A'A$ is symmetric, and its eigenvalues are non-negative.

$$(A'A)v = \lambda v \implies \lambda \|v\|_2^2 = (\lambda v, v) = (A'A v, v) = (Av, Av) = \|Av\|_2^2,$$

$$\implies \lambda = \frac{\|Av\|_2^2}{\|v\|_2^2} \geq 0.$$

Also:

$$\lambda \leq \|A\|_2^2.$$

The 2-norm of a matrix

Whatever A is, $A'A$ is symmetric, and its eigenvalues are non-negative.

Definition

- A **right singular vector** of A is an eigenvector of $A'A$.
- An $s \geq 0$ is a **singular value** of A is $s^2 \in \sigma(A'A)$.

Notation (spectral radius): A square:

$$\rho(A) := \max\{|\lambda| : \lambda \in \sigma(A)\}.$$

So:

$$\|A\|_2 \geq \sqrt{\rho(A'A)}.$$

Characterizing the 2-norm

Theorem: Characterizing the 2-norm

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Proof: We already saw that $\|A\|_2 \geq \sqrt{\rho(A'A)}$.

Now, Let $v \in \mathbb{R}^m$, such that $\|v\|_2 = 1$, and $\|A\|_2 = \|Av\|_2$. Let

$$A'A = QDQ'$$

be the Schur decomposition of $A'A$. Then

$$\begin{aligned}\|A\|_2^2 &= \|Av\|_2^2 = (Av, Av) = (A'Av, v) = \\ &= (QDQ'v, v) = (DQ'v, Q'v).\end{aligned}$$

Denote $w := Q'v$. Since Q' is orthogonal, $\|w\|_2 = \|v\|_2 = 1$.

Characterizing the 2-norm

$$\begin{aligned}\|A\|_2^2 &= \|Av\|_2^2 = (Av, Av) = (A'Av, v) = \\ &= (QDQ'v, v) = (DQ'v, Q'v).\end{aligned}$$

Denote $w := Q'v$. Since Q' is orthogonal, $\|w\|_2 = \|v\|_2 = 1$.
So,

$$\|A\|_2^2 = (Dw, w) = \sum_{i=1}^m D(i, i)w(i)^2 \leq \sum_{i=1}^m \rho(A'A)w(i)^2 = \rho(A'A) \sum_{i=1}^m w(i)^2 = \rho(A'A).$$

So, $\|A\|_2 \leq \sqrt{\rho(A'A)}$.

Demo #2

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Definition of Positive Definiteness