ECEn 671: Mathematics of Signals and Systems

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Section 1

Matrix Norms

Matrix Norms

For matrices $A: \mathbb{C}^m \to \mathbb{C}^n$ we have the following induced norm:

$$\|A\|_{\infty} = \max_{\|x\|_{\infty} = 1} \|Ax\|_{\infty}$$

(Why max not sup?)

Lemma

$$||A||_{\infty} = \max_{i=1:m} \sum_{j=1:n} |a_{ij}|$$

i.e., the largest row sum.

Proof

First show that $||A||_{\infty} \leq \max_{i=1:m} \sum_{i=1:n} |a_{ij}|$:

$$\begin{split} \|A\|_{\infty} &= \max_{\|\mathbf{x}\|_{\infty} = 1} \left\| \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix} \begin{pmatrix} x_{1} \\ \vdots \\ x_{n} \end{pmatrix} \right\|_{\infty} \\ &= \max_{\|\mathbf{x}\|_{\infty} = 1} \left[\max \begin{pmatrix} \left| \sum_{j=1}^{n} a_{1j} x_{j} \right| \\ \vdots \\ \left| \sum_{j=1}^{n} a_{mj} x_{j} \right| \right) \right] \\ &\leq \max_{\mathbf{x} \text{ s.t. } \max_{|\mathbf{x}|} |=1} \left[\max \left(\left| \sum_{j=1}^{n} |a_{1j}| |x_{j}|, \cdots, \sum_{j=1}^{n} |a_{mj}| |x_{j}| \right) \right] \\ &\leq \max_{\|\mathbf{x}\|_{\infty} = 1} \left[\max \left(\left\| \mathbf{x} \right\|_{\infty} \sum_{j=1}^{n} |a_{1j}|, \cdots, \left\| \mathbf{x} \right\|_{\infty} \sum_{j=1}^{m} |a_{mj}| \right) \right] \\ &= \max_{i=1:m} \sum_{j=1}^{m} |a_{ij}| \end{split}$$

Proof, cont.

Now we need to show that $\max_{i=1:m} \sum_{j=1:n} |a_{ij}| \leq ||A||_{\infty}$:

Let $k = \arg\max_{i=1:m} \sum_{j=1:n} |a_{ij}|$ and let \hat{x} be such that

$$\hat{x}_j = egin{cases} 1 & ext{if} & a_{kj} \geq 0 \ -1 & ext{otherwise} \end{cases}$$

then $\|\hat{x}\|_{\infty} = 1$ and then

$$\|A\hat{x}\|_{\infty} = \max_{i=1:m} \sum_{i=1:n} |a_{ij}| \le \max_{\|x\|_{\infty}=1} \|Ax\|_{\infty} = \|A\|_{\infty}.$$

Other Matrix Norms

Lemma

$$\|A\|_1 = \max_{\|x\|_1=1} \|Ax\|_1$$

$$= \max_{j=1:n} \sum_{i=1}^m |a_{ij}| \text{ (largest column sum)}$$

Lemma

$$||A||_2 = \max_i \sqrt{\lambda_i(A^H A)} = largest singular value of A$$

More discussion of this in Chapter 7.



Norm of A^{-1}

Theorem

For induced matrix norms, where A^{-1} exists we have

$$||A^{-1}|| = \frac{1}{\min\limits_{x \neq 0} \frac{||Ax||}{||x||}} = \frac{1}{\min\limits_{||x|| = 1} ||Ax||}$$

Proof.

Let $Ax = b \Rightarrow x = A^{-1}b$ then

$$\begin{aligned} \left\| A^{-1} \right\| &= \max_{b \neq 0} \frac{\left\| A^{-1} b \right\|}{\left\| b \right\|} = \max_{x \neq 0} \frac{\left\| x \right\|}{\left\| A x \right\|} = \max_{x \neq 0} \frac{1}{\frac{\left\| A x \right\|}{\left\| x \right\|}} \\ &= \frac{1}{\min\limits_{x \neq 0} \frac{\left\| A x \right\|}{\left\| x \right\|}} = \frac{1}{\min\limits_{\left\| x \right\| = 1} \left\| A x \right\|} \end{aligned}$$

Frobenius Norm

Definition

The Frobenius norm of a matrix is given by

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

= $\sqrt{tr(A^H A)}$

Fact: The Frobenius norm is NOT an induced norm.

Matrix Convergence

For matrices: convergence in any norm implies convergence in any other norm. In particular

$$\begin{aligned} \|A\|_{2} &\leq \|A\|_{F} \leq \sqrt{n} \, \|A\|_{2} \\ \max |a_{ij}| &\leq \|A\|_{2} \leq \sqrt{mn} \max |a_{ij}| \\ \frac{1}{\sqrt{n}} \, \|A\|_{\infty} &\leq \|A\|_{2} \leq \sqrt{m} \, \|A\|_{\infty} \\ \frac{1}{\sqrt{m}} \, \|A\|_{1} &\leq \|A\|_{2} \leq \sqrt{n} \, \|A\|_{1} \end{aligned}$$