

# CS 591, Fall 2014

## Assignment 1

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- (1) Recall from class that for a matrix  $A = (a_{jk} \in \mathbb{C}^{m \times N})$ , the operator norm of A from  $l_p$  into  $l_p$  is defined as:  $\|A\|_{p \rightarrow p} := \max_{\|x\|_p=1} \|Ax\|_p$

- (a)  $\|A\|_{1 \rightarrow 1} = \max_{k \in [N]} \sum_{j=1}^m |a_{jk}|$   
proof.

$$\text{If } A = \begin{bmatrix} a_{11} & \dots & a_{1N} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mN} \end{bmatrix}, \text{ and } x = \begin{bmatrix} x_1 \\ \vdots \\ x_N \end{bmatrix}, \text{ then } Ax = \begin{bmatrix} a_{11}x_1 + \dots + a_{1N}x_N \\ \vdots \\ a_{m1}x_1 + \dots + a_{mN}x_N \end{bmatrix}$$

Therefore:

$$\|A\|_{1 \rightarrow 1} = \max_{\|x\|_1=1} \|Ax\| = \max_{\|x\|_1=1} \left\| \begin{bmatrix} a_{11}x_1 + \dots + a_{1N}x_N \\ \vdots \\ a_{m1}x_1 + \dots + a_{mN}x_N \end{bmatrix} \right\|_1$$

$$= \max_{\|x\|_1=1} |a_{11}x_1 + \dots + a_{1N}x_N| + \dots + |a_{m1}x_1 + \dots + a_{mN}x_N|$$

Since this is the max over all  $\|x\|_1 = 1$ , this would happen when  $x_i = 1$  for some index i and 0 for all other indices.

The result would then be  $\sum_{j=1}^m |a_{jk}|$  for an index k. We would have to go through each index k from 1 to N to see which would give the greatest result for the maximum. Then:

$$\max_{\|x\|_1=1} |a_{11}x_1 + \dots + a_{1N}x_N| + \dots + |a_{m1}x_1 + \dots + a_{mN}x_N|$$

$$= \max_{k \in [N]} \sum_{j=1}^m |a_{jk}|$$

$$(b) \|A\|_{\infty \rightarrow \infty} = \max_{j \in [m]} \sum_{k=0}^N |a_{jk}|$$

proof.

Using A, x and Ax as defined in part (a):

$$\|A\|_{\infty \rightarrow \infty} = \max_{\|x\|_{\infty}=1} \|Ax\|_{\infty} = \max_{\|x\|_{\infty}=1} \left\| \begin{bmatrix} a_{11}x_1 + \dots + a_{1N}x_N \\ \vdots \\ a_{m1}x_1 + \dots + a_{mN}x_N \end{bmatrix} \right\|_{\infty}$$

$$= \max_{\|x\|_{\infty}=1} \max(|a_{11}x_1 + \dots + a_{1N}x_N|, \dots, |a_{m1}x_1 + \dots + a_{mN}x_N|)$$

Since this is the max over all  $\|x\|_{\infty} = 1$ , this will happen when we take x to be a vector made up of 1's and -1's.

$$= \max(|a_{11}| + \dots + |a_{1N}|, \dots, |a_{m1}| + \dots + |a_{mN}|)$$

$$= \max(\sum_{k=1}^N |a_{1k}|, \dots, \sum_{k=1}^N |a_{mk}|)$$

So the result is the maximum of the L1 norms of every row:

$$= \max_{j \in [m]} \sum_{k=1}^N |a_{jk}|$$

- (c)  $\|A\|_{2 \rightarrow 2} = \sigma_{\max}(A) = \sqrt{\lambda_{\max}(A^*A)}$  where  $\sigma_{\max}(A)$  denotes the largest singular value of A and  $\sqrt{\lambda_{\max}(A^*A)}$  is the largest eigenvalue of  $A^*A$ .

proof.

Let  $\Gamma = A^*A$ , let  $\lambda_1 \dots \lambda_n$  be the eigenvalues of  $\Gamma$ .

Let  $e_1, \dots, e_n$  be the orthonormal basis of the Euclidian Space.

Let  $v \in C^N$  be  $v = a_1 e_1 + \dots + a_n e_n = \sum_i^n a_i e_i$ . So that:  
 $\|v\|^2 = |\langle v, v \rangle| = |\langle \sum_i^n a_i e_i, \sum_i^n a_i e_i \rangle| = \sum_i^n a_i^2$

Now look at  $\Gamma v$ :

$$\Gamma v = \Gamma(\sum_i^n a_i e_i) = \sum_i^n a_i \Gamma(e_i) = \sum_i^n a_i \lambda_i e_i.$$

Denote the largest  $\lambda_i$  value to be  $\lambda_{\max}$ .

$$\begin{aligned} \text{Now want to show } \|A\| &\leq \sqrt{\lambda_{\max}(A^*A)}, \\ \|Av\|^2 &= \langle Av, Av \rangle \\ &= (Av)^T (Av) = v^T A^T A v = \langle v, A^T A v \rangle \\ &= \langle v, \Gamma v \rangle = \langle \sum_i^n a_i e_i, \sum_i^n \lambda_i a_i e_i \rangle \\ &= \sum_i^n (a_i e_i) (\lambda_i a_i e_i)^T = \sum_i^n a_i e_i e_i^T a_i^T \lambda_i^T \\ &= \sum_i^n a_i a_i^T \lambda_i^T \leq \sum_i^n a_i a_i^T \lambda_{\max} = \lambda_{\max} \|v\|^2 \\ \text{So } \|A\|^2 &\leq \lambda_{\max}. \\ \text{So } \|A\| &\leq \sqrt{\lambda_{\max}(A^*A)} \end{aligned}$$

Now want to show  $\|A\| \geq \sqrt{\lambda_{\max}(A^*A)}$ ,

Consider  $x_0 = e_{\max}$ .  $\|x\| = 1$ . Where  $e_{\max}$  is the eigenvector corresponding to  $\lambda_{\max}$ .

$$\begin{aligned} \langle x, \Gamma x \rangle &= \langle e_{\max}, \Gamma(e_{\max}) \rangle = \langle e_{\max}, \lambda_{j_0} e_{\max} \rangle = |\lambda_{\max}|. \\ \text{But } \langle x_0, \Gamma x_0 \rangle &\leq \|A\|^2 \end{aligned}$$

$$\text{So } \|A\| \geq \sqrt{\lambda_{\max}(A^*A)}.$$

$$\text{If } \|A\| \geq \sqrt{\lambda_{\max}(A^*A)} \text{ and } \|A\| \leq \sqrt{\lambda_{\max}(A^*A)},$$

$$\text{Then } \|A\| = \sqrt{\lambda_{\max}(A^*A)}$$

(2) Given a

- (3) Let  $V$  be a normed vector space with an inner product. Prove Cauchy Shqartz Inequality.  
Proof.

Let  $u, v \in V$ . If  $v = 0$  then the equality holds trivially. So assume  $v \neq 0$ .

Look at its orthogonal decomposition.

$$\begin{aligned} u &= \frac{\langle u, v \rangle}{\|v\|^2} v + \left( u - \frac{\langle u, v \rangle}{\|v\|^2} v \right) \\ &= \frac{\langle u, v \rangle}{\|v\|^2} v + w \text{ here } w \text{ is orthognal to } v. \end{aligned}$$

Using Pythagorean Theorem:

$$\begin{aligned} \|u\|^2 &= \left\| \frac{\langle u, v \rangle}{\|v\|^2} v \right\|^2 + \|w\|^2 \\ &\geq \frac{|\langle u, v \rangle|^2}{\|v\|^2}. \end{aligned}$$

$$\text{So: } \|u\|^2 \geq \frac{|\langle u, v \rangle|^2}{\|v\|^2}$$

$$\text{Then: } \|v\|^2 \|u\|^2 \geq |\langle u, v \rangle|^2$$

$$\text{Then: } \|v\| \|u\| \geq |\langle u, v \rangle|$$