

Prefer on Norms (iso. 19-22) - 4/9/19

1. A norm is a function  $\|\cdot\|: \mathbb{R}^n \rightarrow \mathbb{R}$  s.t.

i)  $\|x\| \geq 0 \forall x \in \mathbb{R}^n$

ii)  $\|x\| = 0 \iff x = \vec{0}$

iii)  $\| \lambda x \| = |\lambda| \|x\| \quad \forall x \in \mathbb{R}^n, \lambda \in \mathbb{R}$

iv)  $\|x+y\| \leq \|x\| + \|y\| \quad \forall x, y \in \mathbb{R}^n$

2. The 1-norm is a norm on  $\mathbb{R}^n$  and is defined as

$$\|\cdot\|_1: f(x) = \sum_{i=1}^n |x_i|$$

3. The  $\infty$ -norm is a norm on  $\mathbb{R}^n$  and is defined as

$$\|\cdot\|_\infty: f(x) = \max_{i,j \in n} |x_j|$$

4. The 2-norm is a norm on  $\mathbb{R}^n$  and is defined as

$$\|\cdot\|_2: f(x) = \sqrt{x^T x} = \sqrt{\sum_{i=1}^n |x_i|^2} = \langle x, x \rangle$$

5. Cauchy-Schwarz:  $\forall x, y \in \mathbb{R}^n$

$$|x^T y| \leq \|x\|_2 \|y\|_2$$

6. A real inner product is a function  $\langle \cdot, \cdot \rangle: \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$

where  $\forall x, y, z \in \mathbb{R}^n, \alpha, \beta \in \mathbb{R}$  are such that

i)  $\langle \alpha x + \beta y, z \rangle = \alpha \langle x, z \rangle + \beta \langle y, z \rangle$

ii)  $\langle x, y \rangle = \langle y, x \rangle$

iii)  $\langle x, x \rangle \geq 0$

iv)  $\langle x, x \rangle = 0 \iff x = \vec{0}$

7. A matrix  $A$  is symmetric if  $A^T = A$ .

8. A symmetric matrix is positive definite if

$$x^T A x > 0 \quad \forall x \in \mathbb{R}^n$$

and equality holds only if  $x = \vec{0}$ .

9.  $\mathbb{Z}$   $\|\cdot\|: \mathbb{R}^n \rightarrow \mathbb{R}$  is a norm on vectors, the induced norm of a matrix is

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

and will always be a norm where the domain is  $\mathbb{R}^{n \times n}$ .

10. If  $\|\cdot\|$  is an induced norm on matrices,  $\|I\| = 1$ .

11. For an induced matrix norm  $\|\cdot\|$ ,

$$\|AB\| \leq \|A\| \cdot \|B\|$$

12. The function  $\|\cdot\|: \mathbb{R}^{n \times n} \rightarrow \mathbb{R}$  via

$$\|x\|_A^2 = x^T A x$$

is a norm if  $A$  is symmetric and positive definite.

13. Let  $\lambda_j$  be the eigenvalues of  $A^T A$ . Then

$$\|A\|_2^2 = \max_j \lambda_j$$

14. If  $\|\cdot\|$  is an induced norm on matrices

for which  $\|A\| < 1$  then  $(I - A)^{-1} = \sum_{k=0}^{\infty} A^k$ .