

CHAPTER 7

ITERATIVE TECHNIQUES IN LINEAR ALGEBRA.

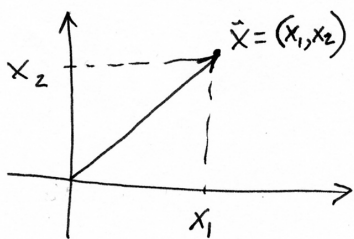
Def - Vector norm is a function

$$\|\cdot\|: \mathbb{R}^n \rightarrow \mathbb{R}$$

With the following properties:

- i) $\|\vec{x}\| \geq 0, \forall \vec{x} \in \mathbb{R}^n$.
- ii) $\|\vec{x}\| = 0 \Leftrightarrow \vec{x} = \vec{0}$.
- iii) $\|\alpha \vec{x}\| = |\alpha| \|\vec{x}\|, \forall \alpha \in \mathbb{R} \text{ and } \vec{x} \in \mathbb{R}^n$.
- iv) $\|\vec{x} + \vec{y}\| \leq \|\vec{x}\| + \|\vec{y}\|, \forall \vec{x}, \vec{y} \in \mathbb{R}^n$.

Example 1 - In $\mathbb{R}^2$ (plane), euclidean norm or distance.



$$\|\vec{x}\| = \sqrt{x_1^2 + x_2^2}$$

magnitude of the vector
distance from origin to the
point $\vec{x}$.

Example 2 - The ℓ_2 norm is defined in $\mathbb{R}^n$
as

$$\|\vec{x}\|_2 = \left(\sum_{i=1}^n x_i^2 \right)^{1/2}$$

Example 3.- ∞ norm of infinite norm

$$\|\vec{x}\|_{\infty} = \max_{1 \leq i \leq n} |x_i|$$

All these definitions are motivated for the necessity to measure errors in the numerical approximations.

The magnitude of the error will depend on the norm being used. In fact,

$\vec{x} = (2, 3, 5)$ is such that

$$\|\vec{x}\|_2 = \sqrt{4+9+25} = \sqrt{38}, \quad \text{while}$$

$$\|\vec{x}\|_{\infty} = \max(|2|, |3|, |5|) = 5.$$

An important theorem. Cauchy-Schwarz inequality

$$(\vec{x}, \vec{y}) \leq \|\vec{x}\| \|\vec{y}\|, \quad \text{or} \quad \sum_{i=1}^n x_i y_i \leq \|\vec{x}\| \|\vec{y}\|$$

Corollary.- For any $\vec{x}, \vec{y} \in \mathbb{R}^n$.

$$\begin{aligned} \|\vec{x} + \vec{y}\|_2 &\leq \|\vec{x}\|_2 + \|\vec{y}\|_2 \\ \text{Proof:- } \|\vec{x} + \vec{y}\|_2^2 &= \sum_{i=1}^n (x_i + y_i)^2 = \sum_{i=1}^n x_i^2 + 2 \sum_{i=1}^n x_i y_i + \sum_{i=1}^n y_i^2 \\ &\leq \|\vec{x}\|_2^2 + 2 \|\vec{x}\|_2 \|\vec{y}\|_2 + \|\vec{y}\|_2^2 = (\|\vec{x}\|_2 + \|\vec{y}\|_2)^2 \end{aligned}$$

$$\Rightarrow \text{TRIANGULAR INEQUALITY: } \|\vec{x} + \vec{y}\| \leq \|\vec{x}\| + \|\vec{y}\|.$$

Definition. - Distance between vectors in $\mathbb{R}^n$.

$$\vec{x} = (x_1, \dots, x_n), \quad \vec{y} = (y_1, \dots, y_n).$$

$$\|\vec{x} - \vec{y}\|_2 = \left[\sum_{i=1}^n (x_i - y_i)^2 \right]^{1/2} \text{ and } \|\vec{x} - \vec{y}\|_\infty = \max_{1 \leq i \leq n} |x_i - y_i|$$

Theorem. - For each $\vec{x} \in \mathbb{R}^n$.

$$\|\vec{x}\|_\infty \leq \|\vec{x}\|_2 \leq \sqrt{n} \|\vec{x}\|_\infty.$$

Proof. -

$$\|\vec{x}\|_\infty^2 = x_j^2 \leq \sum_{i=1}^n x_i^2 = \|\vec{x}\|_2^2 \Rightarrow \boxed{\|\vec{x}\|_\infty^2 \leq \|\vec{x}\|_2^2} \quad (3.1)$$

where $x_j = \max_{1 \leq i \leq n} |x_i| = \|\vec{x}\|_\infty.$

Also,

$$\|\vec{x}\|_2^2 = \sum_{i=1}^n x_i^2 \leq \sum_{i=1}^n x_j^2 = n x_j^2 = n \max_{1 \leq i \leq n} |x_i|^2 = n \|\vec{x}\|_\infty^2.$$

$$\Rightarrow \boxed{\|\vec{x}\|_2 \leq \sqrt{n} \|\vec{x}\|_\infty} \quad (3.2)$$

Combining (3.1) and (3.2).

$$\boxed{\|\vec{x}\|_\infty \leq \|\vec{x}\|_2 \leq \sqrt{n} \|\vec{x}\|_\infty} \quad \checkmark$$

Definition:- Consider the set of all matrices

$$\{A : A \text{ } n \times n \text{ matrix}\} = M_{n \times n}$$

The function

$$\|\cdot\|: M_{n \times n} \rightarrow \mathbb{R}$$

satisfying the following properties for any matrices

$A, B \in M_{n \times n}$ and $\alpha \in \mathbb{R}$ is called a matrix

norm

$$i) \|A\| \geq 0$$

$$ii) \|A\| = 0 \Leftrightarrow A = [0] \text{ matrix with all entries equal to zero.}$$

$$iii) \|\alpha A\| \leq |\alpha| \|A\|$$

$$iv) \|A+B\| \leq \|A\| + \|B\|$$

$$v) \|AB\| \leq \|A\| \|B\|.$$

Thm.:- (Natural or induced matrix norm)

If $\|\cdot\|$ is an arbitrary norm on $\mathbb{R}^n$, then

$$\|A\| \equiv \max_{\|x\|=1} \|A(x)\|$$

is a matrix norm.

Proof:- It's easy.

An alternative representation is given by choosing $\vec{z} \neq \vec{0}$ any vector and considering

$\vec{x} = \vec{z}/\|\vec{z}\|$ unit vector. All unit vectors can be obtained from some $\vec{z} \neq \vec{0}$

Then

$$\max_{\|\vec{x}\|=1} |A\vec{x}| = \max_{\|\vec{z}\| \neq 0} \left\| A \left(\frac{\vec{z}}{\|\vec{z}\|} \right) \right\| =$$

$$= \max_{\vec{z} \neq \vec{0}} \frac{\|A\vec{z}\|}{\|\vec{z}\|}$$

Thus,

$$\boxed{\|A\| = \max_{\vec{z} \neq \vec{0}} \frac{\|A\vec{z}\|}{\|\vec{z}\|}} \quad (5.1)$$

Corollary. For any $\vec{z} \neq \vec{0}$ and any natural norm $\|\cdot\|$

$$\|A\vec{z}\| \leq \|A\| \|\vec{z}\|.$$

Proof. -

For any $\vec{z} \neq \vec{0}$, (5.1) implies

$$\frac{\|A\vec{z}\|}{\|\vec{z}\|} \leq \max \frac{\|A\vec{z}\|}{\|\vec{z}\|} = \|A\|$$

$$\Rightarrow \|A\vec{z}\| \leq \|A\| \|\vec{z}\|.$$

For our norms l_2 and l_∞ in $\mathbb{R}^n$,
we have two natural matrix norms. They are

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

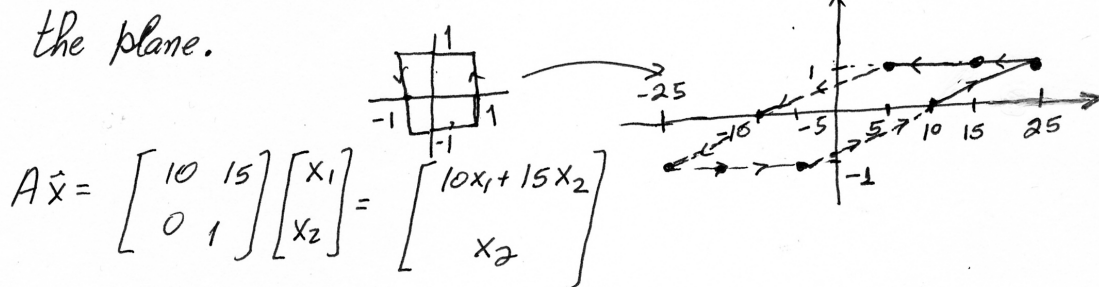
$$\|A\|_2 \equiv \max_{\|x\|_2=1} \|A\tilde{x}\|_2$$

Therefore, the norm of a matrix (as we have defined it)
is the maximum stretch of the unit vectors.

Exercise 4 a) Find $\|A\|_\infty$ of

$$A = \begin{bmatrix} 10 & 15 \\ 0 & 1 \end{bmatrix} \quad \text{if } \tilde{x} \text{ is such that } \|\tilde{x}\|_\infty = 1,$$

then $\tilde{x}$ lies in the boundary of a square ^{of sides 1} in the plane.



$$A\tilde{x} = \begin{bmatrix} 10 & 15 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 10x_1 + 15x_2 \\ x_2 \end{bmatrix}$$

$$\Rightarrow \|A\tilde{x}\|_\infty = \max(|10x_1 + 15x_2|, |x_2|)$$

Obviously, for vectors lying in the square the maximum is reached
when $x_1=1$, $x_2=1$, for the vectors $\tilde{x} = (1, 1)$.

Thus $\|A\|_\infty = \max_{\|\tilde{x}\|=1} \|A\tilde{x}\|_\infty = |10x_1 + 15x_2| = \underline{\underline{25}}$

Thm.- $A = (a_{ij})$ is an $n \times n$ matrix, then

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

Proof. - let's show that first

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

let $\tilde{x}$ be such that $\|\tilde{x}\|_{\infty} = 1$

$$\Rightarrow \|A\tilde{x}\|_{\infty} = \max_{1 \leq i \leq n} |(A\tilde{x})_i| = \max_{1 \leq i \leq n} \left| \sum_{j=1}^n a_{ij} x_j \right| \leq$$

$$\stackrel{\text{Triang.}}{\leq} \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}| |x_j| \leq$$

$$\leq \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}| \left(\max_{1 \leq j \leq n} |x_j| \right) = \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}| \|\tilde{x}\|_{\infty} \stackrel{=1}{=} \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}|$$

$$\therefore \boxed{\|A\|_{\infty} \leq \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}|} \quad (7.1)$$