

تحليل دالي

المحاضرة الثانية



Definition 1.10 : A *metric space* is a pair (X, d) , where X is a set and d is a metric on X (or distance function on X), that is, a function defined on $X \times X$ such that for all $x, y, z \in X$, we have:

- (1) d is real-valued, finite and nonnegative function.
- (2) $d(x, y)=0$ if and only if $x=y$
- (3) $d(x, y) = d(y, x)$ (Symmetry).
- (4) $d(x, y) \leq d(x, z)+d(z, y)$ (Triangle inequality).

Examples (H.W. 2-6)

1) Real line $\mathbb{R}$: this is the set of all real numbers, taken with the usual metric defined by:

$$d(x, y) = |x - y| \quad \forall x, y \in \mathbb{R}$$

Sol.

[1] 1- d is real, finite & $d = |x - y| \geq 0$

2) $d(x, y) = 0 \leftrightarrow |x - y| = 0 \leftrightarrow x - y = 0 \leftrightarrow x = y \quad \forall x, y \in \mathbb{R}$

3) $d(x, y) = |x - y| = |-(y - x)| = |y - x| = d(y, x) \quad \forall x, y \in \mathbb{R}$

4) $d(x, y) = |x - y| = |x - z + z - y| \leq |x - z| + |z - y| = d(x, z) + d(z, y) \quad \forall x, y, z \in \mathbb{R}$

Then $(\mathbb{R}, d)$ is a metric space.

A **norm** on a vector space is a way of measuring distance between vectors.

Definition 1.11.: A **norm** on a linear space V over F is a function $\| \cdot \| : V \rightarrow \mathbb{R}$ with the properties that :

- (1) $\| x \| \geq 0$ & $\| x \| = 0 \leftrightarrow x = 0$ (positive definite);
- (2) $\| x + y \| \leq \| x \| + \| y \|$ for all $x, y \in V$ (triangle inequality);
- (3) $\| \alpha x \| = |\alpha| \| x \|$ for all $x \in V$ and $\alpha \in F$.

In Definition 1.11(3) we are assuming that F is $\mathbb{R}$ or $\mathbb{C}$ and $|\cdot|$ denotes the usual absolute value. If $\| \cdot \|$ is a function with properties (2) and (3) only it is called a **semi-norm**.

Definition 1.12. A **normed linear space** is a linear space V with a norm $\| \cdot \|$ (sometimes we write $\| \cdot \|_V$).

Theorem 1.13. If V is a normed space then:

- 1) $\| 0 \| = 0$
- 2) $\| x \| = \| -x \|$ for every $x \in V$.
- 3) $\| x - y \| = \| y - x \|$ for every $x, y \in V$.
- 4) $|\| x \| - \| y \|| \leq \| x - y \|$ for every $x, y \in V$.

Proof:

Properties (1), (2) and (3) conclude directly from the definition, to prove property (4):

$$x = (x-y)+y$$

$$\|x\| = \|(x-y)+y\| \leq \|x-y\| + \|y\| \rightarrow \|x\| - \|y\| \leq \|x-y\| \dots(1)$$

Similarly:

$$\|y\| - \|x\| \leq \|x-y\|$$

$$-(\|x\| - \|y\|) \leq \|x-y\| \rightarrow (\|x\| - \|y\|) \geq -\|x-y\| \dots(2)$$

From (1) & (2), we get:

$$-\|x-y\| \leq \|x\| - \|y\| \leq \|x-y\| \rightarrow |\|x\| - \|y\|| \leq \|x-y\|$$

Examples 1.14.:- [H.W.6,7]

[1] The vector space V is normed v.s. with the norm $\|x\| = |x|$ for all $x \in V$.

Proof:

1) Since $|x| \geq 0 \rightarrow \|x\| \geq 0$.

2) $\|x\| = 0 \leftrightarrow |x| = 0 \leftrightarrow x = 0$

3) Let $x \in V, \alpha \in F$, then

$$\|\alpha x\| = |\alpha x| = |\alpha| |x| = |\alpha| \|x\|$$

1) Let $x, y \in V$, then:

$$\|x + y\| = |x + y| \leq |x| + |y| = \|x\| + \|y\|$$

[2] Let $V = \mathbb{R}^n$ with the usual Euclidean norm

$$\|x\| = \|x\|_2 = \left(\sum_{j=1}^n |x_j|^2 \right)^{1/2}$$

proof:

1) Since

$$x_j^2 \geq 0 \text{ for all } j = 1, 2, \dots, n \rightarrow \|x\| \geq 0$$

$$2) \|x\| = 0 \Leftrightarrow \left(\sum_{j=1}^n |x_j|^2\right)^{1/2} = 0 \Leftrightarrow \sum_{j=1}^n |x_j|^2 = 0$$

$$\Leftrightarrow x_j^2 = 0 \text{ for all } j = 1, 2, \dots, n \Leftrightarrow x_j = 0 \text{ for all } j = 1, 2, \dots, n \Leftrightarrow x = 0$$

3) Let $x \in \mathbb{R}^n$, $\alpha \in \mathbb{R}$:

$$\alpha x = \alpha (x_1, \dots, x_n) = (\alpha x_1, \dots, \alpha x_n)$$

$$\|\alpha x\| = \left(\sum_{j=1}^n |\alpha x_j|^2\right)^{1/2} = |\alpha| \left(\sum_{j=1}^n |x_j|^2\right)^{1/2} = |\alpha| \|x\|.$$

4) Let $x, y \in \mathbb{R}^n$:

$$x + y = (x_1, \dots, x_n) + (y_1, \dots, y_n) = (x_1 + y_1, \dots, x_n + y_n)$$

$$\|x + y\| = \left(\sum_{j=1}^n |x_j + y_j|^2\right)^{1/2}$$

By using *Minkowski's inequality* where $p=2$, we have:

$$\|x + y\| = \left(\sum_{i=1}^n |x_i + y_i|^2\right)^{1/2} \leq \left(\sum_{i=1}^n |x_i|^2\right)^{1/2} + \left(\sum_{i=1}^n |y_i|^2\right)^{1/2} = \|x\| + \|y\|$$

[3] There are many other norms on $\mathbb{R}^n$, called the p -norms. For $1 \leq p < \infty$ defined by:

$$\|x\|_p = \left(\sum_{j=1}^n |x_j|^p \right)^{1/p}$$

Then $\|\cdot\|_p$ is a norm on V (to check the triangle inequality use *MinKowski's Inequality*)

$$\left(\sum_{j=1}^n |x_j + y_j|^p \right)^{1/p} \leq \left(\sum_{j=1}^n |x_j|^p \right)^{1/p} + \left(\sum_{j=1}^n |y_j|^p \right)^{1/p}$$

[4] There is another norm corresponding to $p = \infty$, defined by:

$$\|x\|_\infty = \max_{1 \leq j \leq n} \{|x_j|\}$$

where $\|\cdot\|: \mathbb{R}^n \rightarrow \mathbb{R}$ and $x = (x_1, \dots, x_n)$.

proof:

1) Since $|x_i| \geq 0$ for all $i=1, \dots, n \rightarrow \|x\| \geq 0$.

2) $\|x\| = 0 \leftrightarrow \max \{|x_1|, \dots, |x_n|\} = 0 \leftrightarrow |x_i| = 0$ for all $i=1, \dots, n$

$\leftrightarrow x_i = 0$ for all $i=1, \dots, n \leftrightarrow x=0$

3) Let $x \in \mathbb{R}^n$ and $\alpha \in \mathbb{R}$, then

$$\alpha x = \alpha (x_1, \dots, x_n) = (\alpha x_1, \dots, \alpha x_n)$$

$$\| \alpha x \| = \max \{ | \alpha x_1 |, \dots, | \alpha x_n | \}$$

$$= \max \{ | \alpha | | x_1 |, \dots, | \alpha | | x_n | \}$$

$$= | \alpha | \max \{ | x_1 |, \dots, | x_n | \}$$

$$= | \alpha | \| x \|$$

4) Let $x, y \in \mathbb{R}^n$

$$x + y = (x_1, \dots, x_n) + (y_1, \dots, y_n) = (x_1 + y_1, \dots, x_n + y_n)$$

$$\| x + y \| = \max \{ | x_1 + y_1 |, \dots, | x_n + y_n | \}$$

$$\leq \max \{ | x_1 | + | y_1 |, \dots, | x_n | + | y_n | \}$$

$$\leq \max \{ | x_1 |, \dots, | x_n | \} + \max \{ | y_1 |, \dots, | y_n | \}$$

$$= \| x \| + \| y \|$$

[5] Let $X = C[a, b]$, and put $\|f\| = \sup_{t \in [a, b]} |f(t)|$. This is called the uniform or supremum norm.

proof:

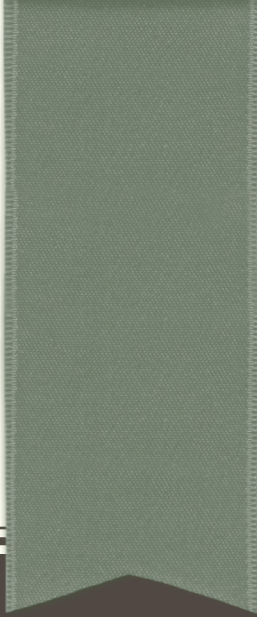
1) Since $|f(t)| \geq 0$ for all $t \in [a, b] \rightarrow \|f\| \geq 0$.

$$\begin{aligned}\|f\| = 0 &\leftrightarrow \sup_{t \in [a, b]} |f(t)| = 0 \leftrightarrow |f(t)| = 0 \text{ for all } t \in [a, b] \\ &\leftrightarrow f(t) = 0 \text{ for all } t \in [a, b] \leftrightarrow f = 0.\end{aligned}$$

2) Let $f \in X$, $\alpha \in \mathbb{R}$, then:

$$\begin{aligned}\|\alpha f\| &= \sup\{| \alpha f(t) |: t \in [a, b]\} \\ &= \sup\{|\alpha| |f(t)| : t \in [a, b]\} \\ &= |\alpha| \sup\{|f(t)| : t \in [a, b]\} \\ &= |\alpha| \|f\|.\end{aligned}$$

$$\begin{aligned}3) \|f + g\| &= \sup\{|(f + g)(t)| : t \in [a, b]\} = \sup\{|f(t) + g(t)| : t \in [a, b]\} \\ &\leq \sup\{|f(t)| + |g(t)| : t \in [a, b]\} \\ &\leq \sup\{|f(t)| : t \in [a, b]\} + \sup\{|g(t)| : t \in [a, b]\} = \|f\| + \|g\|.\end{aligned}$$



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