

**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 vector 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):

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

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

Similarly:

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

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

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]

Let  $V$  be a vector space  $V$  is normed v.s. with the norm  $\|x\| = |x|$  for all  $x \in V$ .

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

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

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

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

4) 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



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

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

4) 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  $x_j^2 \geq 0$  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:

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

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



[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

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

$$\begin{aligned} \|\alpha x\| &= \max \{ |\alpha x_1|, \dots, |\alpha x_n| \} \\ &= \max \{ |\alpha| |x_1|, \dots, |\alpha| |x_n| \} \\ &= |\alpha| \max \{ |x_1|, \dots, |x_n| \} \\ &= |\alpha| \|x\| \end{aligned}$$

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

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

$$\begin{aligned} \|x + y\| &= \max \{ |x_1 + y_1|, \dots, |x_n + y_n| \} \\ &\leq \max \{ |x_1| + |y_1|, \dots, |x_n| + |y_n| \} \\ &\leq \max \{ |x_1|, \dots, |x_n| \} + \max \{ |y_1|, \dots, |y_n| \} \\ &= \|x\| + \|y\| \end{aligned}$$

[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  $t \in [a, b] \rightarrow \|f\| \geq 0$ .

2)  $\|f\| = 0 \leftrightarrow \sup_{t \in [a, b]} |f(t)| = 0 \leftrightarrow |f(t)| = 0$  for all  $t \in [a, b]$

$f(t) = 0$  for all  $t \in [a, b] \leftrightarrow f = 0$ .

Let  $x \in X$ ,  $\alpha \in \mathbb{R}$ , then:

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

4)  $\|f + g\| = \sup \{ |f(t) + g(t)| : t \in [a, b] \} = \sup \{ |f(t)| + |g(t)| : t \in [a, b] \}$   
 $\leq \sup \{ |f(t)| : t \in [a, b] \} + \sup \{ |g(t)| : t \in [a, b] \}$