

# Normed Spaces and Banach Spaces

## Vector Space

**Definition.** A **vector space** (or **linear space**) over a field  $K$  is a nonempty set  $X$  of elements  $x, y, \dots$  (called **vectors**) together with two algebraic operations. These operations are called vector addition and multiplication of vectors by scalars, that is, by elements of  $K$ .

**Vector addition** associates with every ordered pair  $(x, y)$  of vectors a vector  $x + y$ , called the sum of  $x$  and  $y$ , in such a way that the following properties hold! Vector addition is commutative and associative, that is, for all vectors we have

$$\begin{aligned}x + y &= y + x \\x + (y + z) &= (x + y) + z;\end{aligned}$$

furthermore, there exists a vector  $0$ , called the **zero vector**, and for every vector  $x$  there exists a vector  $-x$ , such that for all vectors we have

$$\begin{aligned}x + 0 &= x \\x + (-x) &= 0.\end{aligned}$$

**Multiplication by scalars** associates with every vector  $x$  and scalar  $\alpha$  a vector  $\alpha x$ , called the product of  $\alpha$  and  $x$ , in such a way that for all vectors  $x, y$  and scalars  $\alpha, b$  we have

$$\begin{aligned}\alpha(\beta x) &= (\alpha\beta)x \\1x &= x\end{aligned}$$

and the distributive laws

$$\begin{aligned}\alpha(x + y) &= \alpha x + \alpha y \\(\alpha + \beta)x &= \alpha x + \beta x.\end{aligned}$$

## Normed Space, Banach Space

**Definition.** A **normed space**  $X$  is a vector space with a norm defined on it. A **Banach space** is a complete normed space (complete in the metric defined by the norm). Here a **norm** on a vector space  $X$  is a real-valued function on  $X$  whose value at an  $x \in X$  is denoted by  $\|x\|$  and which has the properties:

$$(N1) \quad \|x\| \geq 0.$$

$$(N2) \quad \|x\| = 0 \quad \Leftrightarrow \quad x = 0.$$

$$(N3) \quad \|\alpha x\| = |\alpha| \|x\|.$$

$$(N4) \quad \|x + y\| \leq \|x\| + \|y\| \quad (\text{Triangle inequality}).$$

A norm on  $X$  naturally defines a metric  $d$  on  $X$  which is given by

$$d(x, y) = \|x - y\|$$

and is called the **metric induced by the norm**. The normed space just defined is denoted by  $(X, \|\cdot\|)$  or simply by  $X$ .

Note that

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

And this implies that: *The norm is continuous, i.e.  $x \mapsto \|x\|$  is a continuous mapping of  $(X, \|\cdot\|)$  into  $\mathbb{R}$ .*

A metric  $d$  induced by a norm satisfies translation invariance:

$$\begin{aligned}d(x + a, y + a) &= d(x, y) \\d(\alpha x, \alpha y) &= |\alpha| d(x, y).\end{aligned}$$

Here are some useful spaces:

1. Euclidean space  $\mathbb{R}^n$  and unitary space  $C^n$  are Banach spaces with norm defined by

$$\|x\| = \left(\sum_{j=1}^n |\xi_j|^2\right)^{1/2} = \sqrt{|\xi_1|^2 + \cdots + |\xi_n|^2}.$$

2. Space  $\ell^p$  is a Banach space with norm defined by

$$\|x\| = \left(\sum_{j=1}^{\infty} |\xi_j|^p\right)^{1/p}.$$

3. Space  $\ell^\infty$  is a Banach space with norm defined by

$$\|x\| = \sup_j |\xi_j|.$$

4. Space  $C[a, b]$  is a Banach space with norm defined by

$$\|x\| = \max_{t \in [a, b]} |x(t)|,$$

but is incomplete with norm defined by

$$\|x\| = \int_0^1 |x(t)| dt.$$

5. The vector space of all continuous real-valued functions on  $[a, b]$  forms a normed space  $X$  with norm defined by

$$\|x\| = \left(\int_a^b x(t)^2 dt\right)^{1/2}.$$

*This space is not complete.* The space  $X$  can be completed by the theorem introduced in previous chapter. The completion is denoted by  $L^2[a, b]$ . This is a Banach space.

More generally, for any fixed real number  $p \geq 1$ , the Banach space  $L^p[a, b]$  is the completion of the normed space which consists of all continuous real-valued functions on  $[a, b]$ , as before, and the norm defined by

$$\|x\|_p = \left(\int_a^b |x(t)|^p dt\right)^{1/p}.$$

**Definition.** A **subspace**  $Y$  of a normed space  $X$  is a subspace of  $X$  considered as a vector space, with the norm obtained by restricting the norm on  $X$  to the subset  $Y$ . This norm on  $Y$  is said to be induced by the norm on  $X$ . If  $Y$  is closed in  $X$ , then  $Y$  is called a **closed subspace** of  $X$ .

**Theorem.** A subspace  $y$  of a Banach space  $X$  is complete if and only if the set  $Y$  is closed in  $X$ .

In a normed space we may go an important step further and use series as follows.

If  $(x_k)$  is a sequence in a normed space  $X$ , we can associate with  $(x_k)$  the sequence  $(s_n)$  of **partial sums**

$$s_n = x_1 + x_2 + \cdots + x_n$$

where  $n = 1, 2, \dots$ . If  $(s_n)$  is convergent, say,

$$s_n \rightarrow s, \quad \text{that is,} \quad \|s_n - s\| \rightarrow 0,$$

then the **infinite series** or, briefly, **series**

$$\sum_{k=1}^{\infty} x_k = x_1 + x_2 + \cdots$$

is said to **converge** or to be **convergent**,  $s$  is called the sum of the series and we write

$$s = \sum_{k=1}^{\infty} x_k = x_1 + x_2 + \cdots$$

If  $\|x_1\| + \|x_2\| + \cdots$  converges, the above series is said to be **absolutely convergent**.

*In a normed space  $X$ , absolute convergence implies convergence if and only if  $X$  is complete.*

**Definition.** If a normed space  $X$  contains a sequence  $(e_n)$  with the property that for every  $x \in X$  there is a unique sequence of scalars  $(\alpha_n)$  such that

$$\|x - (\alpha_1 e_1 + \cdots + \alpha_n e_n)\| \rightarrow 0 \quad (\text{as } n \rightarrow \infty)$$

then  $(e_n)$  is called a **Schauder basis** (or **basis**) for  $X$ . The series  $\sum_{k=1}^{\infty} \alpha_k e_k$  which has the sum  $x$  is then called the **expansion** of  $x$  with respect to  $(e_n)$ , and we write

$$x = \sum_{k=1}^{\infty} \alpha_k e_k.$$

For example,  $\ell^p$  has a basis  $e_n = (\delta_{nj})$ . Thus,

$$e_1 = (1, 0, 0, 0, \dots)$$

$$e_2 = (0, 1, 0, 0, \dots)$$

$$e_3 = (0, 0, 1, 0, \dots)$$

etc.

If a normed space  $X$  has a Schauder basis, then  $X$  is separable. But conversely, a separable normed space may not have a Schauder basis according to the counterexample constructed by P. Enflo (1973).

**Theorem.** Let  $X = (X, \|\cdot\|)$  be a normed space. Then there is a Banach space  $\hat{X}$  and an isometry  $A$  from  $X$  onto a subspace  $W$  of  $\hat{X}$  which is dense in  $\hat{X}$ . The space  $\hat{X}$  is unique, except for isometries.

## Finite Dimension

**Lemma (Bound for linear combinations).** Let  $\{x_1, \dots, x_n\}$  be a linearly independent set of vectors in a normed space  $X$ . Then there is a number  $c > 0$  such that for every choice of scalars  $\alpha_1, \dots, \alpha_n$ , we have

$$\|\alpha_1 x_1 + \cdots + \alpha_n x_n\| \geq c(|\alpha_1| + \cdots + |\alpha_n|) \quad (c > 0).$$

**Theorem.** Every finite dimensional subspace  $Y$  of a normed space  $X$  is complete. In particular, every finite dimensional normed space is complete.

*Sketch of Proof.* Consider the basis of  $Y$  and make use of the lemma (bound for linear combinations). ■

**Theorem.** Every finite dimensional subspace  $Y$  of a normed space  $X$  is closed in  $X$ .

**Definition.** A norm  $\|\cdot\|$  on a vector space  $X$  is said to be **equivalent** to a norm  $\|\cdot\|_0$  on  $X$  if there are positive numbers  $a$  and  $b$  such that for all  $x \in X$  we have

$$a\|x\|_0 \leq \|x\| \leq b\|x\|_0.$$

In fact, equivalent norms on  $X$  define the same topology for  $X$ .

**Theorem.** On a finite dimensional vector space  $X$ , any norm  $\|\cdot\|$  is equivalent to any other norm  $\|\cdot\|_0$ .

*Sketch of Proof.* Consider the basis of  $X$  and make use of the lemma (bound of linear combinations). ■

This theorem is of considerable practical importance. For instance, it implies that convergence or divergence of a sequence in a finite dimensional vector space does not depend on the particular choice of a norm on that space.

## Compactness

**Definition.** A metric space  $X$  is said to be **compact** if every sequence in  $X$  has a convergent subsequence. A subset  $M$  of  $X$  is said to be **compact** if  $M$  is compact considered as a subspace of  $X$ , that is, if every sequence in  $M$  has a convergent subsequence whose limit is an element of  $M$ .

**Lemma.** A compact subset  $M$  of a metric space is closed and bounded.

The converse of this lemma is in general false.

However, for a finite dimensional normed space, we do have:

**Theorem.** *In a finite dimensional normed space  $X$ , any subset  $M \subset X$  is compact if and only if  $M$  is closed and bounded.*

*Sketch of the Proof.* Compactness implies closedness and boundedness by the above lemma. For the converse, consider the basis of  $X$  and make use of the lemma (bound of linear combinations). ■

**F. Riesz's Lemma.** *Let  $Y$  and  $Z$  be subspaces of a normed space  $X$  (of any dimension), and suppose that  $Y$  is closed and is a proper subset of  $Z$ . Then for every real number  $\theta$  in the interval  $(0, 1)$  there is a  $z \in Z$  such that*

$$\|z\| = 1, \quad \|z - y\| \geq \theta \text{ for all } y \in Y.$$

*Proof.* We consider any  $v \in Z - Y$  and denote its distance from  $Y$  by  $a$ , i.e.

$$a = \inf_{y \in Y} \|v - y\|.$$

Clearly,  $a > 0$  since  $Y$  is closed. We now take any  $\theta \in (0, 1)$ . By the definition of an infimum there is a  $y_0 \in Y$  such that

$$a \leq \|v - y_0\| \leq \frac{a}{\theta}.$$

Let

$$z = c(v - y_0) \quad \text{where} \quad c = \frac{1}{\|v - y_0\|}.$$

Then  $\|z\| = 1$ , and we show that

$$\begin{aligned} \|z - y\| &= \|c(v - y_0) - y\| \\ &= c\|v - y_0 - c^{-1}y\| \\ &= c\|v - y_1\| \end{aligned}$$

where  $y_1 = y_0 + c^{-1}y \in Y$ .

Thus

$$\|z - y\| = c\|v - y_1\| \geq ca = \frac{a}{\|v - y_0\|} \geq \frac{a}{a/\theta} = \theta.$$

Since  $y \in Y$  was arbitrary, this completes the proof. ■

**Theorem.** *If a normed space  $X$  has the property that the closed unit ball  $M = \{x \mid \|x\| \leq 1\}$  is compact, then  $X$  is finite dimensional.*

*Proof.* We assume that  $M$  is compact but  $\dim X = \infty$ , and show that this leads to a contradiction. We choose any  $x_1$  of norm 1. This  $x_1$  generates a one dimensional subspace  $X_1$  of  $X$ , which is closed and is a proper subspace of  $X$ . By Riesz's lemma, there is a  $x_2 \in X$  of norm 1 such that

$$\|x_2 - x_1\| \geq \theta = \frac{1}{2}.$$

The elements  $x_1, x_2$  generate a two dimensional proper closed subspace  $X_2$  of  $X$ . By Riesz's lemma there is an  $x_3$  of norm 1 such that for all  $x \in X_2$  we have

$$\|x_3 - x\| \geq \frac{1}{2}.$$

In particular,

$$\|x_3 - x_1\| \geq \frac{1}{2},$$

$$\|x_3 - x_2\| \geq \frac{1}{2}.$$

Proceeding by induction, we obtain a sequence  $(x_n)$  of elements  $x_n \in M$  such that

$$\|x_m - x_n\| \geq \frac{1}{2} \quad (m \neq n).$$

Obviously,  $(x_n)$  cannot have a convergent subsequence. This contradicts the compactness of  $M$ . Hence our assumption  $\dim X = \infty$  is false, and  $\dim X < \infty$ . ■

Compact sets are important since they are "well-behaved": they have several basic properties similar to those of finite sets and not shared by noncompact sets. In connection with continuous mappings a fundamental property is that compact sets have compact images, as follows.

**Theorem.** *Let  $X$  and  $Y$  be metric spaces and  $T : X \rightarrow Y$  a continuous mapping. Then the image of a compact subset  $M$  of  $X$  under  $T$  is compact.*

**Theorem.** *A continuous mapping  $T$  of a compact subset  $M$  of a metric space  $X$  into  $R$  assumes a maximum and a minimum at some points of  $M$ .*

## Linear Operators

**Definition.** A linear operator  $T$  is an operator such that

- (i) the domain  $\mathcal{D}(T)$  of  $T$  is a vector space and the range  $\mathcal{R}(T)$  lies in a vector space over the same field.
- (ii) for all  $x, y \in \mathcal{D}(T)$  and scalars  $\alpha$ ,

$$T(x + y) = Tx + Ty,$$

$$T(\alpha x) = \alpha Tx.$$

Observe the notation: we write  $Tx$  instead of  $T(x)$ ;  $\mathcal{D}(T)$  denotes the domain of  $T$ ;  $\mathcal{R}(T)$  denotes the range of  $T$ ;  $\mathcal{N}(T)$  denotes the **null space** of  $T$  (the set of all  $x \in \mathcal{D}(T)$  such that  $Tx = 0$ ).

We shall now consider some basic examples of linear operators:

1. **Identity operator.** The identity operator  $I_X : X \rightarrow X$  is defined by  $I_X x = x$  for all  $x \in X$ . We also write simply  $I$  for  $I_X$ ; thus,  $Ix = x$ .
2. **Zero operator.** The zero operator  $\mathbf{0} : X \rightarrow Y$  is defined by  $\mathbf{0}x = 0$  for all  $x \in X$ .
3. **Differentiation.** Let  $X$  be the vector space of all polynomials on  $[a, b]$ . We may define a linear operator  $T$  on  $X$  by setting

$$Tx(t) = x'(t)$$

for every  $x \in X$ , where the prime denotes the differentiation with respect to  $t$ . This operator  $T$  maps  $X$  onto itself.

4. **Integration.** A linear operator  $T$  from  $C[a, b]$  into itself can be defined by

$$Tx(t) = \int_a^t x(\tau) d\tau \quad t \in [a, b].$$

5. **Multiplication by  $t$ .** A linear operator  $T$  from  $C[a, b]$  into itself is defined by

$$Tx(t) = tx(t).$$

6. **Elementary vector algebra.** The cross product with one factor kept fixed defines a linear operator  $T_1 : R^3 \rightarrow R^3$ . Similarly, the dot product with one fixed factor defines a linear operator  $T_2 : R^3 \rightarrow R$ , say,

$$T_2 x = x \cdot a = \xi_1 \alpha_1 + \xi_2 \alpha_2 + \xi_3 \alpha_3$$

where  $a = (\alpha_j) \in R^3$  is fixed.

7. **Matrices.** A real matrix  $A = (\alpha_{jk})$  with  $r$  rows and  $n$  columns defines an operator  $T : R^n \rightarrow R^r$  by means of

$$y = Ax$$

where  $x = (\xi_j)$  has  $n$  components and  $y = (\eta_i)$  has  $r$  components and both vectors are written as column vectors because of the usual convention of matrix multiplication.

If  $A$  were complex, it would define a linear operator from  $C^n$  into  $C^r$ .

**Theorem.** Let  $T$  be a linear operator. Then  $\mathcal{R}(T)$  is a vector space; if  $\dim \mathcal{L}(T) = n < \infty$ , then  $\dim \mathcal{R}(T) \leq n$ ; the null space  $\mathcal{N}(T)$  is a vector space.

A important consequence of the second claim is that: *linear operators preserve linear dependence.*

**Theorem (Inverse operator).** Let  $T$  be a linear operator. The inverse  $T^{-1} : \mathcal{R}(T) \rightarrow \mathcal{D}(T)$  exists if and only if

$$Tx = 0 \quad \Rightarrow \quad x = 0.$$

If  $T^{-1}$  exists, it is a linear operator. Moreover, if  $T^{-1}$  exists and  $\dim \mathcal{D}(T) = n < \infty$ , then  $\dim \mathcal{R}(T) = \dim \mathcal{D}(T)$ .

**Theorem.** Let  $T : X \rightarrow Y$  and  $S : Y \rightarrow Z$  be bijective linear operators, where  $X, Y, Z$  are vector spaces. Then the inverse  $(ST)^{-1} : Z \rightarrow X$  of the product  $ST$  exists, and

$$(ST)^{-1} = T^{-1}S^{-1}.$$

**Definition.** Let  $X$  and  $Y$  be normed spaces and  $T : \mathcal{D}(T) \rightarrow Y$  a linear operator, where  $\mathcal{D}(T) \subset X$ . The operator  $T$  is said to be **bounded** if there is a real number  $c$  such that for all  $x \in \mathcal{L}(T)$ ,

$$\|Tx\| \leq c\|x\|.$$

**Definition.** Let  $T$  be a bounded linear operator. Then the **norm** of  $T$  is defined as

$$\|T\| = \sup_{\substack{x \in \mathcal{D}(T) \\ \|x\| \neq 0}} \frac{\|Tx\|}{\|x\|}.$$

It's clear that the norm defined by the above formula satisfies (N1) to (N4).

An alternative formula for the norm of  $T$  is

$$\|T\| = \sup_{\substack{x \in \mathcal{D}(T) \\ \|x\|=1}} \|Tx\|.$$

Let's take a look at some typical examples:

1. **Identity operator.** The identity operator  $I : X \rightarrow X$  on a normed space  $X \neq \{0\}$  is bounded and has norm  $\|I\| = 1$ .
2. **Zero operator.** The zero operator  $\mathbf{0} : X \rightarrow Y$  on a normed space  $X$  is bounded and has norm  $\|\mathbf{0}\| = 0$ .
3. **Differentiation operator.** Let  $X$  be the normed space of all polynomials on  $J = [0, 1]$  with norm given  $\|x\| = \max |x(t)|$ ,  $t \in J$ . A differentiation operator  $T$  is defined on  $X$  by

$$Tx(t) = x'(t).$$

This operator is linear but not bounded. Indeed, let  $x_n(t) = t^n$ , where  $n \in \mathbb{N}$ . Then  $\|x_n\| = 1$  and

$$Tx_n(t) = x'_n(t) = nt^{n-1}$$

so that  $\|Tx_n\| = n$  and  $\|Tx_n\|/\|x_n\| = n$ . Since  $n \in \mathbb{N}$  is arbitrary, this shows that there is no fixed number  $c$  such that  $\|Tx_n\|/\|x_n\| \leq c$ .

4. **Integral operator.** We can define an integral operator  $T : C[0, 1] \rightarrow C[0, 1]$  by

$$y = Tx \quad \text{where} \quad y(t) = \int_0^1 k(t, \tau)x(\tau)d\tau.$$

Here  $k$  is a given function, which is called the *kernel* of  $T$  and is assumed to be continuous on the closed square  $G = J \times J$  in the  $t\tau$ -plane, where  $J = [0, 1]$ . This operator is linear and bounded.

5. **Matrix.** A real matrix  $A = (\alpha_{jk})$  with  $r$  rows and  $n$  columns defines an operator  $T : R^n \rightarrow R^r$  by means of

$$y = Ax.$$

Then  $T$  is bounded.

**Theorem.** If a normed space  $X$  is finite dimensional, then every linear operator on  $X$  is bounded.

*Sketch of Proof.* Consider the basis of  $Y$  and make use of the lemma (bound for linear combinations). ■

**Theorem.** Let  $T : \mathcal{D}(T) \longrightarrow Y$  be a linear operator, where  $\mathcal{D}(T) \subset X$  and  $X, Y$  are normed spaces. Then:

(a)  $T$  is continuous if and only if  $T$  is bounded.

(b) If  $T$  is continuous at a single point, it is continuous.

**Corollary.** Let  $T$  be a bounded linear operator. Then:

(a)  $x_n \longrightarrow x$  [where  $x_n, x \in \mathcal{D}(T)$ ] implies  $Tx_n \longrightarrow Tx$ .

(b) The null space  $\mathcal{N}(T)$  is closed.

The **restriction** of an operator  $T : \mathcal{D}(T) \longrightarrow Y$  to a subset  $B \subset \mathcal{D}(T)$  is denoted by  $T|_B$  and is the operator defined by

$$T|_B : B \longrightarrow Y, \quad T|_B x = Tx \text{ for all } x \in B.$$

An **extension** of  $T$  to a set  $M \supset \mathcal{D}(T)$  is an operator

$$\tilde{T} : M \rightarrow Y \quad \text{such that} \quad \tilde{T}|_{\mathcal{D}(T)} = T,$$

that is,  $\tilde{T}x = Tx$  for all  $x \in \mathcal{D}(T)$ .

**Theorem.** Let  $T : \mathcal{D}(T) \longrightarrow Y$  be a bounded linear operator, where  $\mathcal{D}(T)$  lies in a normed space  $X$  and  $Y$  is a Banach space. Then  $T$  has an extension

$$\tilde{T} : \overline{\mathcal{D}(T)} \rightarrow Y$$

where  $\tilde{T}$  is a bounded linear operator of norm

$$\|\tilde{T}\| = \|T\|.$$