

## UNIT I

### Riemann Stieltjes integral

#### LESSON-1

#### DEFINITION AND EXISTENCE OF THE INTEGRAL

##### 1.0 Introduction

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##### 1.0 Introduction

In UG , you have studied the definition of Riemann integral and the existence theorems on it. Riemann Stieltjes integral is the extension of Riemann integral. In particular , if we set  $(x)=x$  in Riemann Stieltjes integral, we get the Riemann integral.

##### 1.1 Aims and objectives

After studying this lesson you should be able to

- Find whether the Riemann Stieltjes integral of a function exists
- Find the Riemann Stieltjes integral.

##### 1.2 Definition of Riemann –Stieltjes integral and existence theorems.

###### Definition

Let  $[a, b]$  be a given interval. A partition  $P$  of  $[a, b]$  is a finite set of points  $x_0, x_1, x_2, \dots, x_n$  such that

$$a = x_0 \leq x_1 \leq x_2 \leq \dots \leq x_n = b .$$

###### Definition

Let  $f$  be a monotonically increasing function on  $[a, b]$ .

Corresponding to any partition  $P$  of  $[a, b]$ ,

$$\alpha_i = (x_i) - (x_{i-1}), \quad i=1,2,\dots,n.$$

Then  $\alpha_i \geq 0$ .

Let  $f$  be a bounded real valued function on  $[a, b]$ .

$$\text{Let } U(P, f, \alpha) = \sum_{i=1}^n M_i \Delta \alpha_i$$

$$L(P, f, \alpha) = \sum_{i=1}^n m_i \Delta \alpha_i$$

where  $M_i = \text{Sup}\{ f(x) / x \in [x_{i-1}, x_i] \}$

and  $m_i = \text{inf}\{ f(x) / x \in [x_{i-1}, x_i] \}$

We define the upper Riemann-Stieltjes integral of  $f$  as

$$\int_a^b f d\alpha = \inf U(P, f, \alpha)$$

and the lower Riemann –Stieltjes integral of  $f$  as

$$\int_a^b \mathbf{f} d\alpha = \sup \mathbf{L}(\mathbf{P}, \mathbf{f}, \alpha),$$

where the infimum and supremum are taken over all partitions  $\mathbf{P}$  of  $[a, b]$ .

$$\text{If } \int_a^b \mathbf{f} d\alpha = \overline{\int_a^b \mathbf{f} d\alpha},$$

their common value is denoted by  $\int_a^b \mathbf{f} d\alpha$  or  $\int_a^b \mathbf{f}(\mathbf{x}) d\alpha(\mathbf{x})$ .

This is called the Riemann-Stieltjes integral of  $\mathbf{f}$  with respect to  $\alpha$  on  $[a, b]$ .

If  $\int_a^b \mathbf{f} d\alpha$  exists, then  $\mathbf{f}$  is said to be integrable with respect to  $\alpha$  on  $[a, b]$ .

It is written as  $\mathbf{f} \in \mathbf{R}(\alpha)$  on  $[a, b]$ .

### Definition

A partition  $\mathbf{P}^*$  is said to be a refinement of  $\mathbf{P}$ , if  $\mathbf{P}^* \supseteq \mathbf{P}$ .

Remark:

Given two partitions  $\mathbf{P}_1$  and  $\mathbf{P}_2$  of  $[a, b]$ , their common refinement is  $\mathbf{P}^* = \mathbf{P}_1 \cup \mathbf{P}_2$ .

### Theorem 1.1

If  $\mathbf{P}^*$  is a refinement of  $\mathbf{P}$ , then

$$\mathbf{U}(\mathbf{P}^*, \mathbf{f}, \alpha) \leq \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha)$$

and  $\mathbf{L}(\mathbf{P}^*, \mathbf{f}, \alpha) \geq \mathbf{L}(\mathbf{P}, \mathbf{f}, \alpha)$ .

### Proof:

Assume that  $\mathbf{P}^*$  contains just one point more than  $\mathbf{P}$ .

Let this be  $c$  and  $x_{i-1} < c < x_i$ .

$$\text{Let } M_i' = \sup\{f(x) / x \in [x_{i-1}, c]\}$$

$$\text{and } M_i'' = \sup\{f(x) / x \in [c, x_i]\}.$$

Then  $M_i' \leq M_i$  and  $M_i'' \leq M_i$ .

$$\text{Consider } \mathbf{U}(\mathbf{P}^*, \mathbf{f}, \alpha) = \sum_{\substack{k=1 \\ k \neq i}}^n \mathbf{M}_k \Delta\alpha_k + M_i' [\alpha(c) - \alpha(x_{i-1})] + M_i'' [\alpha(x_i) - \alpha(c)]$$

$$\leq \sum_{\substack{k=1 \\ k \neq i}}^n \mathbf{M}_k \Delta\alpha_k + M_i [\alpha(c) - \alpha(x_{i-1})] + M_i [\alpha(x_i) - \alpha(c)]$$

$$\leq \sum_{\substack{k=1 \\ k \neq i}}^n \mathbf{M}_k \Delta\alpha_k + M_i [\alpha(x_i) - \alpha(x_{i-1})]$$

$$\leq \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha).$$

Similarly we can prove that

$$\mathbf{L}(\mathbf{P}^*, \mathbf{f}, \alpha) \geq \mathbf{L}(\mathbf{P}, \mathbf{f}, \alpha).$$

Hence the theorem.

**Theorem 1.2**

$$\underline{\int_a^b f d\alpha} \leq \overline{\int_a^b f d\alpha}.$$

**Proof:**

Let  $P_1$  and  $P_2$  be any partitions of  $[a, b]$ .

Let  $P^* = P_1 \cup P_2$ .

Then  $P^*$  is the common refinement of  $P_1$  as well as  $P_2$ .

Therefore by theorem 1.1,

$$U(P^*, f, \alpha) \leq U(P_1, f, \alpha) \quad \dots(1)$$

$$\text{and } L(P^*, f, \alpha) \geq L(P_2, f, \alpha) \quad \dots(2)$$

Also we know that

$$L(P^*, f, \alpha) \leq U(P^*, f, \alpha) \quad \dots(3)$$

From (1), (2) and (3), we get

$$L(P_2, f, \alpha) \leq L(P^*, f, \alpha) \leq U(P^*, f, \alpha) \leq U(P_1, f, \alpha)$$

Therefore for any two partitions  $P_1$  and  $P_2$  of  $[a, b]$ , we have

$$L(P_2, f, \alpha) \leq U(P_1, f, \alpha).$$

Keeping  $P_2$  fixed and varying  $P_1$  over all partitions of  $[a, b]$ ,

$$L(P_2, f, \alpha) \leq \inf U(P, f, \alpha)$$

Now this is true for all partitions  $P_2$  of  $[a, b]$ .

Therefore ,

$$\sup L(P, f, \alpha) \leq \inf U(P, f, \alpha).$$

Therefore

$$\underline{\int_a^b f d\alpha} \leq \overline{\int_a^b f d\alpha}.$$

Hence the theorem.

**Theorem 1.3**

$f \in R(\alpha)$  on  $[a, b]$  if and only if there exists a partition  $P$  of  $[a, b]$  such that

$$U(P, f, \alpha) - L(P, f, \alpha) < \varepsilon.$$

**Proof:**

Let  $f \in R(\alpha)$  on  $[a, b]$  .

$$\text{Then } \underline{\int_a^b f d\alpha} = \overline{\int_a^b f d\alpha}, \quad \dots(1)$$

$$\text{where } \underline{\int_a^b f d\alpha} = \inf U(P, f, \alpha)$$

$$\text{and } \overline{\int_a^b f d\alpha} = \sup L(P, f, \alpha),$$

Therefore , by definition of infimum and supremum,

for given  $\varepsilon > 0$ , there exists a partition  $P_1$  of  $[a, b]$  such that

$$U(P_1, f, \alpha) < \overline{\int_a^b f d\alpha} + \varepsilon/2 \quad \dots(2)$$

and a partition  $P_2$  of  $[a, b]$  such that

$$L(P_2, f, \alpha) > \underline{\int_a^b f d\alpha} - \varepsilon/2. \quad \dots(3)$$

Let  $P = P_1 \cup P_2$ .

Then by theorem 1.1,

$$U(P, f, \alpha) \leq U(P_1, f, \alpha) \quad \dots(4)$$

$$\text{and } \mathbf{L}(\mathbf{P}, \mathbf{f}, \alpha) \geq \mathbf{L}(\mathbf{P}_2, \mathbf{f}, \alpha) \quad \dots(5)$$

Therefore ,

From (1), (2), (3), (4) and (5), we get

$$\begin{aligned} \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha) &\leq \mathbf{U}(\mathbf{P}_1, \mathbf{f}, \alpha) \\ &< \overline{\int_a^b \mathbf{f}d\alpha} + \varepsilon/2 \\ &< \int_a^b \mathbf{f}d\alpha + \varepsilon/2 \\ &< \mathbf{L}(\mathbf{P}_2, \mathbf{f}, \alpha) + \varepsilon/2 + \varepsilon/2 \\ &< \mathbf{L}(\mathbf{P}, \mathbf{f}, \alpha) + \varepsilon \end{aligned}$$

Therefore there exists a partition P of [a, b] such that

$$\mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha) - \mathbf{L}(\mathbf{P}, \mathbf{f}, \alpha) < \varepsilon .$$

Conversely ,

assume that there exists a partition P of [a, b] such that

$$\mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha) - \mathbf{L}(\mathbf{P}, \mathbf{f}, \alpha) < \varepsilon . \quad \dots(6)$$

For every partition P of [a, b] , we have

$$\mathbf{L}(\mathbf{P}, \mathbf{f}, \alpha) \leq \overline{\int_a^b \mathbf{f}d\alpha} \leq \int_a^b \mathbf{f}d\alpha \leq \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha) \quad \dots(7)$$

From (6) and (7), we get that

$$0 \leq \overline{\int_a^b \mathbf{f}d\alpha} - \int_a^b \mathbf{f}d\alpha \leq \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha) - \mathbf{L}(\mathbf{P}, \mathbf{f}, \alpha) < \varepsilon .$$

This is true for every  $\varepsilon > 0$ .

$$\text{Hence } \overline{\int_a^b \mathbf{f}d\alpha} - \int_a^b \mathbf{f}d\alpha = 0 .$$

Therefore,

$$\overline{\int_a^b \mathbf{f}d\alpha} = \int_a^b \mathbf{f}d\alpha .$$

Hence  $\mathbf{f} \in \mathbf{R}(\alpha)$  on [a, b].

Hence the theorem.

#### Theorem1.4

If  $\mathbf{f}$  is continuous on [a, b], then  $\mathbf{f} \in \mathbf{R}(\alpha)$  on [a, b].

**Proof:**

Let  $\varepsilon > 0$ .

Choose  $\eta > 0$  such that  $[\alpha(b) - \alpha(a)] \eta < \varepsilon$ .

Since  $\mathbf{f}$  is continuous on [a, b] and [a, b] is compact,  $\mathbf{f}$  is uniformly continuous on [a, b]. Therefore for this  $\eta > 0$ , there exists a  $\delta > 0$  such that

$$|f(x) - f(t)| < \eta \text{ whenever } x, t \in [a, b] \text{ with } |x-t| < \delta . \quad \dots(1)$$

If P is any partition of [a, b] such that  $\Delta x_i < \delta$ ,

$$\begin{aligned} \text{then } M_i - m_i &= \sup \{ |f(x) - f(t)| / x, t \in [x_{i-1}, x_i] \} \\ &\leq \eta, \quad i=1, 2, \dots, n. \end{aligned}$$

$$\text{Therefore } \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha) - \mathbf{L}(\mathbf{P}, \mathbf{f}, \alpha) = \sum_{i=1}^n M_i \Delta \alpha_i - \sum_{i=1}^n m_i \Delta \alpha_i$$

$$\begin{aligned}
&= \sum_{i=1}^n (M_i - m_i) \Delta \alpha_i \\
&\leq \eta \sum_{i=1}^n \Delta \alpha_i \\
&\leq \eta [\alpha(b) - \alpha(a)] \\
&< \varepsilon.
\end{aligned}$$

Therefore  $f \in \mathbf{R}(\alpha)$  on  $[a, b]$ .

Hence the theorem.

### Theorem 1.5

If  $f$  is monotonic on  $[a, b]$ , and if  $\alpha$  is continuous on  $[a, b]$ , then  $f \in \mathbf{R}(\alpha)$  on  $[a, b]$ .

#### Proof:

Let  $\alpha$  be increasing on  $[a, b]$ .

Let  $\varepsilon > 0$  be given.

Choose  $n$  large enough such that

$$[\alpha(b) - \alpha(a)]/n < \varepsilon/[f(b) - f(a)].$$

Choose a partition  $P$  such that  $\Delta \alpha_i = [\alpha(b) - \alpha(a)]/n$ .

Let  $f$  be increasing on  $[a, b]$ .

Hence  $f(x_{i-1}) \leq f(x) \leq f(x_i)$  whenever  $x_{i-1} \leq x \leq x_i$ .

Therefore,  $M_i = f(x_i)$  and  $m_i = f(x_{i-1})$ ,  $i=1, 2, \dots, n$ .

$$\begin{aligned}
\text{Therefore } \quad \mathbf{U}(P, f, \alpha) - \mathbf{L}(P, f, \alpha) &= \sum_{i=1}^n M_i \Delta \alpha_i - \sum_{i=1}^n m_i \Delta \alpha_i \\
&= \sum_{i=1}^n (M_i - m_i) \Delta \alpha_i \\
&= \sum_{i=1}^n (f(x_i) - f(x_{i-1})) [(\alpha(b) - \alpha(a))/n] \\
&= [(\alpha(b) - \alpha(a))/n] [f(b) - f(a)] \\
&< \varepsilon
\end{aligned}$$

Therefore  $f \in \mathbf{R}(\alpha)$  on  $[a, b]$

Hence the theorem.

### Check your progress

1. Suppose  $P_1 = \{0, 1/4, 1/2, 3/4, 1\}$  and  $P_2 = \{0, 1/3, 2/3, 1\}$

Is  $P_1$  a refinement of  $P_2$ ?

2. Can you find a refinement that is finer than both  $P_1$  and  $P_2$ ?

3. Can you prove that  $\int_a^b d\alpha(x) = \alpha(b) - \alpha(a)$ , directly from the definition?

4. If  $f(x) = 0$  for all irrational  $x$ ,  $f(x) = 1$  for all rational  $x$ , is  $f \in \mathbf{R}$  on  $[a, b]$  for any  $a < b$ .

### 1.3 Let us sum up

In this lesson we have seen the definition of Riemann Stieltjes integral, lower Riemann Stieltjes integral and the upper Riemann Stieltjes integral and that the

- Lower integral is always less than or equal to upper integral

And the necessary conditions for the existence of the integral are

- $f$  is continuous on  $[a, b]$
- $f$  is monotonic on  $[a, b]$ , and if  $\alpha$  is continuous on  $[a, b]$

and the necessary and sufficient condition is that

- there exists a partition  $P$  of  $[a, b]$  such that  $U(P, f, \alpha) - L(P, f, \alpha) < \varepsilon$ .

#### 1.4 Lesson End Activities

1. Suppose  $f \geq 0$ ,  $f$  is continuous on  $[a, b]$  and  $\int_a^b f(x) dx = 0$ . Prove that  $f(x) = 0 \forall x \in [a, b]$ .
2. Suppose  $\alpha$  is an increasing function on  $[a, b]$  and  $\alpha$  is continuous at  $x_0 \in [a, b]$  &  $f : [a, b] \rightarrow \mathbb{R}$  is defined by  $f(x_0) = 1$  &  $f(x) = 0 \forall x \neq x_0$ . Prove that  $f \in R(\alpha)$  and  $\int f d\alpha = 0$ .
3. Give an Example of a function in which  $\int f d\alpha < \overline{\int} f d\alpha$

#### 1.5 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Wiley and Sons, New York, 1976.
2. W. Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

**LESSON - 2****PROPERTIES OF INTEGRALS****2.0 Introduction****2.1 Aims and objectives****2.2 Properties of integrals****2.3 Let us sum up****2.4 Lesson End Activities****2.5 References****2.0 Introduction**

In this lesson ,we are going to see the properties of Riemann Stieltjes integral.

**2.1 Aims and objectives**

After studying this lesson you should be able to

- Identify the properties of the integral and
- use them to find the Riemann stieltjes integral of functions

**2.2 Properties of integral.****Theorem 2.1**

If  $f_1 \in \mathbf{R}(\alpha)$  and  $f_2 \in \mathbf{R}(\alpha)$  on  $[a, b]$ , then  $f_1 + f_2 \in \mathbf{R}(\alpha)$  and

$$\int_a^b (f_1 + f_2) d\alpha = \int_a^b f_1 d\alpha + \int_a^b f_2 d\alpha$$

**Proof:**

Let  $f = f_1 + f_2$  and  $P$  be any partition of  $[a, b]$ .

Consider

$$U(P, f, \alpha) = \sum_{k=1}^n M_k \Delta \alpha_k$$

$$\begin{aligned} \text{where } M_k &= \sup\{f(x) / x \in [x_{k-1}, x_k]\} \\ &= \sup\{f_1(x) + f_2(x) / x \in [x_{k-1}, x_k]\} \\ &\leq \sup\{f_1(x) / x \in [x_{k-1}, x_k]\} + \sup\{f_2(x) / x \in [x_{k-1}, x_k]\} \\ &\leq M_k' + M_k'' \end{aligned}$$

Therefore ,

$$\sum_{k=1}^n M_k \Delta \alpha_k \leq \sum_{k=1}^n M_k' \Delta \alpha_k + \sum_{k=1}^n M_k'' \Delta \alpha_k$$

Therefore,

$$U(P, f, \alpha) \leq U(P, f_1, \alpha) + U(P, f_2, \alpha) \quad \dots(1)$$

$$\text{Similarly, } L(P, f, \alpha) \geq L(P, f_1, \alpha) + L(P, f_2, \alpha) \quad \dots(2)$$

Since  $f_1 \in \mathbf{R}(\alpha)$  on  $[a, b]$  and  $f_2 \in \mathbf{R}(\alpha)$  on  $[a, b]$ ,

for given  $\varepsilon > 0$ , there exists partitions  $P_1$  and  $P_2$  such that

$$U(P_1, f_1, \alpha) - L(P_1, f_1, \alpha) < \varepsilon / 2 \quad \dots(3)$$

$$\text{and } U(P_2, f_2, \alpha) - L(P_2, f_2, \alpha) < \varepsilon / 2 \quad \dots(4)$$

Let  $P = P_1 \cup P_2$

Then we know that

$$U(P, f_1, \alpha) \leq U(P_1, f_1, \alpha)$$

$$\text{and } U(\mathbf{P}, \mathbf{f}_2, \alpha) \leq U(\mathbf{P}_2, \mathbf{f}_2, \alpha)$$

$$\text{and } L(\mathbf{P}, \mathbf{f}_1, \alpha) \geq L(\mathbf{P}_1, \mathbf{f}_1, \alpha)$$

$$\text{and } L(\mathbf{P}, \mathbf{f}_2, \alpha) \geq L(\mathbf{P}_2, \mathbf{f}_2, \alpha).$$

Therefore ,

$$U(\mathbf{P}, \mathbf{f}_1, \alpha) - L(\mathbf{P}, \mathbf{f}_1, \alpha) \leq U(\mathbf{P}_1, \mathbf{f}_1, \alpha) - L(\mathbf{P}_1, \mathbf{f}_1, \alpha) < \varepsilon/2 \quad \dots(5)$$

$$\text{and } U(\mathbf{P}, \mathbf{f}_2, \alpha) - L(\mathbf{P}, \mathbf{f}_2, \alpha) \leq U(\mathbf{P}_2, \mathbf{f}_2, \alpha) - L(\mathbf{P}_2, \mathbf{f}_2, \alpha) < \varepsilon/2 \quad \dots(6)$$

From (1), (2), (5) and (6), we get

$$\begin{aligned} U(\mathbf{P}, \mathbf{f}, \alpha) - L(\mathbf{P}, \mathbf{f}, \alpha) &\leq [U(\mathbf{P}, \mathbf{f}_1, \alpha) + U(\mathbf{P}, \mathbf{f}_2, \alpha)] - [L(\mathbf{P}, \mathbf{f}_1, \alpha) + L(\mathbf{P}, \mathbf{f}_2, \alpha)] \\ &\leq [U(\mathbf{P}, \mathbf{f}_1, \alpha) - L(\mathbf{P}, \mathbf{f}_1, \alpha)] + [U(\mathbf{P}, \mathbf{f}_2, \alpha) - L(\mathbf{P}, \mathbf{f}_2, \alpha)] \\ &< \varepsilon/2 + \varepsilon/2 \\ &< \varepsilon. \end{aligned}$$

Hence  $\mathbf{f} = \mathbf{f}_1 + \mathbf{f}_2 \in \mathbf{R}(\alpha)$  on  $[a, b]$ .

For the same Partition  $\mathbf{P}$  of  $[a, b]$ ,

$$\begin{aligned} U(\mathbf{P}, \mathbf{f}_1, \alpha) &< L(\mathbf{P}, \mathbf{f}_1, \alpha) + \varepsilon/2 \\ &< \sup\{L(\mathbf{P}, \mathbf{f}_1, \alpha)\} + \varepsilon/2 \\ &< \int_a^b \mathbf{f}_1 d\alpha + \varepsilon/2 \\ &< \int_a^b \mathbf{f}_1 d\alpha + \varepsilon/2 \end{aligned}$$

[Since  $\mathbf{f}_1 \in \mathbf{R}(\alpha)$  on  $[a, b]$ ,  $\int_a^b \mathbf{f}_1 d\alpha = \overline{\int_a^b \mathbf{f}_1 d\alpha} = \underline{\int_a^b \mathbf{f}_1 d\alpha}$  .]

Similarly  $U(\mathbf{P}, \mathbf{f}_2, \alpha) < \int_a^b \mathbf{f}_2 d\alpha + \varepsilon/2$

Therefore,

$$\begin{aligned} \int_a^b \mathbf{f} d\alpha &= \inf\{U(\mathbf{P}, \mathbf{f}, \alpha)\} \\ &< U(\mathbf{P}, \mathbf{f}, \alpha) \\ &< U(\mathbf{P}, \mathbf{f}_1, \alpha) + U(\mathbf{P}, \mathbf{f}_2, \alpha) \\ &< \int_a^b \mathbf{f}_1 d\alpha + \varepsilon/2 + \int_a^b \mathbf{f}_2 d\alpha + \varepsilon/2 \\ &< \int_a^b \mathbf{f}_1 d\alpha + \int_a^b \mathbf{f}_2 d\alpha + \varepsilon \end{aligned}$$

$$\text{Hence } \int_a^b \mathbf{f} d\alpha \leq \int_a^b \mathbf{f}_1 d\alpha + \int_a^b \mathbf{f}_2 d\alpha \quad \dots(7)$$

Replacing  $\mathbf{f}_1$  by  $-\mathbf{f}_1$  and  $\mathbf{f}_2$  by  $-\mathbf{f}_2$ , we get

$$-\int_a^b \mathbf{f} d\alpha \leq -\int_a^b \mathbf{f}_1 d\alpha - \int_a^b \mathbf{f}_2 d\alpha$$

Multiplying both sides by (-1), we get

$$\int_a^b \mathbf{f} d\alpha \geq \int_a^b \mathbf{f}_1 d\alpha + \int_a^b \mathbf{f}_2 d\alpha \quad \dots(8)$$

From (7) and (8), we get that

$$\int_a^b f d\alpha = \int_a^b f_1 d\alpha + \int_a^b f_2 d\alpha$$

Hence the theorem.

### Theorem 2.2

If  $f \in \mathbf{R}(\alpha)$  on  $[a, b]$ ,

then  $cf \in \mathbf{R}(\alpha)$  on  $[a, b]$ , for any constant  $c$  and

$$\int_a^b cf d\alpha = c \int_a^b f d\alpha.$$

**Proof:**

If  $c = 0$ , then the result is true.

Assume that  $c > 0$ .

Since  $f \in \mathbf{R}(\alpha)$  on  $[a, b]$ ,

for given  $\varepsilon > 0$ , there exists a partition  $P$  of  $[a, b]$  such that

$$U(P, f, \alpha) - L(P, f, \alpha) < \varepsilon / c.$$

Consider

$$U(cP, f, \alpha) - L(cP, f, \alpha) = \sum_{k=1}^n M_k' \Delta\alpha_k - \sum_{k=1}^n m_k' \Delta\alpha_k$$

$$\begin{aligned} \text{where } M_k' &= \sup\{(cf)(x) / x \in [x_{k-1}, x_k]\} \\ &= c \sup\{f(x) / x \in [x_{k-1}, x_k]\} \\ &= cM_k \end{aligned}$$

Similarly  $m_k' = cm_k$

Therefore,

$$U(cP, f, \alpha) = c \sum_{k=1}^n M_k \Delta\alpha_k = c U(P, f, \alpha)$$

$$\text{and } L(cP, f, \alpha) = c \sum_{k=1}^n m_k \Delta\alpha_k = c L(P, f, \alpha)$$

Therefore  $U(cP, f, \alpha) - L(cP, f, \alpha) = c[U(P, f, \alpha) - L(P, f, \alpha)] < \varepsilon$ .

Hence  $cf \in \mathbf{R}(\alpha)$  on  $[a, b]$ .

Therefore for the same  $P$ ,

$$\begin{aligned} U(cP, f, \alpha) &< L(cP, f, \alpha) + \varepsilon \\ &\leq \sup L(cP, f, \alpha) + \varepsilon \\ &\leq c \sup L(P, f, \alpha) + \varepsilon \\ &\leq c \int_a^b f d\alpha + \varepsilon \end{aligned}$$

Therefore,

$$\begin{aligned} \inf U(cP, f, \alpha) &\leq U(cP, f, \alpha) \\ &\leq c \int_a^b f d\alpha + \varepsilon \end{aligned}$$

Therefore,

$$\int_a^b cf d\alpha \leq c \int_a^b f d\alpha \quad \dots(1)$$

Replacing  $f$  by  $-f$ , we get

$$-\int_a^b cf d\alpha \leq c[-\int_a^b f d\alpha]$$

Multiplying both sides by  $(-1)$ , we get

$$\int_a^b c f d\alpha \geq c \int_a^b f d\alpha \quad \dots(2)$$

From (1) and (2),

$$\int_a^b c f d\alpha = c \int_a^b f d\alpha .$$

Hence the theorem.

### Theorem 2.3

If  $f_1 \in \mathbf{R}(\alpha)$  on  $[a, b]$ ,  $f_2 \in \mathbf{R}(\alpha)$  on  $[a, b]$  and

$f_1(x) \leq f_2(x)$  on  $[a, b]$ ,

then  $\int_a^b f_1 d\alpha \leq \int_a^b f_2 d\alpha$

**Proof:**

Let  $P$  be any partition of  $[a, b]$ .

Since  $f_1(x) \leq f_2(x)$ ,

$$\sup\{f_1(x) / x \in [x_{k-1}, x_k]\} \leq \sup\{f_2(x) / x \in [x_{k-1}, x_k]\}$$

Therefore ,

$$U(P, f_1, \alpha) \leq U(P, f_2, \alpha)$$

Therefore,

$$\inf U(P, f_1, \alpha) \leq \inf U(P, f_2, \alpha)$$

Therefore,

$$\overline{\int_a^b f_1 d\alpha} \leq \overline{\int_a^b f_2 d\alpha}$$

Therefore ,

$$\int_a^b f_1 d\alpha \leq \int_a^b f_2 d\alpha$$

[since  $f_1 \in \mathbf{R}(\alpha)$  on  $[a, b]$ ,  $f_2 \in \mathbf{R}(\alpha)$  on  $[a, b]$

$$\int_a^b f_1 d\alpha = \overline{\int_a^b f_1 d\alpha} \quad \text{and} \quad \int_a^b f_2 d\alpha = \overline{\int_a^b f_2 d\alpha} ] .$$

Hence the theorem.

### Theorem 2.4

If  $f \in \mathbf{R}(\alpha)$  on  $[a, b]$  and if  $a < c < b$ , then  $f \in \mathbf{R}(\alpha)$  on  $[a, c]$  and  $f \in \mathbf{R}(\alpha)$  on  $[c, b]$ , and

$$\int_a^c f d\alpha + \int_c^b f d\alpha = \int_a^b f d\alpha .$$

**Proof:**

Since  $f \in \mathbf{R}(\alpha)$  on  $[a, b]$ , for given  $\varepsilon > 0$ , there exists a partition  $P$  of  $[a, b]$  such that

$$U(P, f, \alpha) - L(P, f, \alpha) < \varepsilon .$$

Let  $P_1 = P \cap [a, c]$  and  $P_2 = P \cap [c, b]$

The  $P_1$  is a partition of  $[a, c]$  and  $P_2$  is a partition of  $[c, b]$ .

Also on  $[a, c]$ ,

$$U(P_1, f, \alpha) - L(P_1, f, \alpha) < U(P, f, \alpha) - L(P, f, \alpha) < \varepsilon .$$

and on  $[c, b]$ ,

$$U(P_2, f, \alpha) - L(P_2, f, \alpha) < U(P, f, \alpha) - L(P, f, \alpha) < \varepsilon$$

Therefore,

$$f \in \mathbf{R}(\alpha) \text{ on } [a, c] \text{ and } [c, b].$$

For any partition  $P$  of  $[a, b]$ ,

$$\text{Since } P = P_1 \cup P_2,$$

$$\begin{aligned}
U(\mathbf{P}, \mathbf{f}, \alpha) &= U(\mathbf{P}_1, \mathbf{f}, \alpha) + U(\mathbf{P}_2, \mathbf{f}, \alpha) \\
&\geq \inf U(\mathbf{P}_1, \mathbf{f}, \alpha) + \inf U(\mathbf{P}_2, \mathbf{f}, \alpha) \\
&= \overline{\int_a^c \mathbf{f} d\alpha} + \overline{\int_c^b \mathbf{f} d\alpha}
\end{aligned}$$

Therefore,

$$\inf U(\mathbf{P}, \mathbf{f}, \alpha) \geq \overline{\int_a^c \mathbf{f} d\alpha} + \overline{\int_c^b \mathbf{f} d\alpha}$$

Therefore,

$$\overline{\int_a^b \mathbf{f} d\alpha} \geq \overline{\int_a^c \mathbf{f} d\alpha} + \overline{\int_c^b \mathbf{f} d\alpha}$$

Therefore ,

$$\int_a^b \mathbf{f} d\alpha \geq \int_a^c \mathbf{f} d\alpha + \int_c^b \mathbf{f} d\alpha \quad \dots(1)$$

Similarly using lower sums,

$$\begin{aligned}
L(\mathbf{P}, \mathbf{f}, \alpha) &= L(\mathbf{P}_1, \mathbf{f}, \alpha) + L(\mathbf{P}_2, \mathbf{f}, \alpha) \\
&\leq \sup L(\mathbf{P}_1, \mathbf{f}, \alpha) + \sup L(\mathbf{P}_2, \mathbf{f}, \alpha) \\
&\leq \underline{\int_a^c \mathbf{f} d\alpha} + \underline{\int_c^b \mathbf{f} d\alpha}
\end{aligned}$$

Therefore ,

$$\sup L(\mathbf{P}, \mathbf{f}, \alpha) \leq \underline{\int_a^c \mathbf{f} d\alpha} + \underline{\int_c^b \mathbf{f} d\alpha}$$

Therefore ,

$$\underline{\int_a^b \mathbf{f} d\alpha} \leq \underline{\int_a^c \mathbf{f} d\alpha} + \underline{\int_c^b \mathbf{f} d\alpha}$$

Therefore ,

$$\int_a^b \mathbf{f} d\alpha \leq \int_a^c \mathbf{f} d\alpha + \int_c^b \mathbf{f} d\alpha \quad \dots(2)$$

From (1) and (2), we get

$$\int_a^b \mathbf{f} d\alpha = \int_a^c \mathbf{f} d\alpha + \int_c^b \mathbf{f} d\alpha .$$

Hence the theorem.

### Theorem2.5

If  $\mathbf{f} \in \mathbf{R}(\alpha)$  on  $[a, b]$  and if  $|f(x)| \leq M$  on  $[a, b]$ ,

$$\text{then } \left| \int_a^b \mathbf{f} d\alpha \right| \leq M[\alpha(\mathbf{b}) - \alpha(\mathbf{a})]$$

#### Proof:

Let P be any partition of  $[a, b]$ ,

Since  $|f(x)| \leq M$  on  $[a, b]$ ,

$$\begin{aligned}
M_k &= \sup \{f(x) / x \in [x_{k-1}, x_k]\} \\
&\leq M, \text{ for all } k=1, 2, \dots, n.
\end{aligned}$$

Since  $\mathbf{f} \in \mathbf{R}(\alpha)$  on  $[a, b]$ ,

$$\begin{aligned}
\int_a^b \mathbf{f} d\alpha &= \overline{\int_a^b \mathbf{f} d\alpha} = \inf U(\mathbf{P}, \mathbf{f}, \alpha) \\
&\leq U(\mathbf{P}, \mathbf{f}, \alpha)
\end{aligned}$$

$$\begin{aligned}
 &= \sum_{k=1}^n \mathbf{M}_k \Delta \alpha_k \\
 &\leq M \sum_{k=1}^n \Delta \alpha_k \\
 &= M[\alpha(\mathbf{b}) - \alpha(\mathbf{a})] \qquad \dots(1)
 \end{aligned}$$

Replacing  $f$  by  $-f$ , we get

$$- \int_a^b \mathbf{f} d\alpha \leq M[\alpha(\mathbf{b}) - \alpha(\mathbf{a})] \qquad \dots(2)$$

From (1) and (2),

$$\left| \int_a^b \mathbf{f} d\alpha \right| \leq M[\alpha(\mathbf{b}) - \alpha(\mathbf{a})]$$

Hence the theorem.

**Theorem 2.6**

If  $\mathbf{f} \in \mathbf{R}(\alpha_1)$  on  $[a, b]$  and  $\mathbf{f} \in \mathbf{R}(\alpha_2)$  on  $[a, b]$ ,

then  $\mathbf{f} \in \mathbf{R}(\alpha_1 + \alpha_2)$  on  $[a, b]$  and

$$\int_a^b \mathbf{f} d(\alpha_1 + \alpha_2) = \int_a^b \mathbf{f} d\alpha_1 + \int_a^b \mathbf{f} d\alpha_2$$

**Proof:**

Let  $P$  be any partition of  $[a, b]$ .

Let  $\alpha = \alpha_1 + \alpha_2$

$$\begin{aligned}
 \text{Then } \Delta \alpha_k &= (\alpha_1 + \alpha_2)(\mathbf{x}_k) - (\alpha_1 + \alpha_2)(\mathbf{x}_{k-1}) \\
 &= \alpha_1(\mathbf{x}_k) + \alpha_2(\mathbf{x}_k) - [\alpha_1(\mathbf{x}_{k-1}) + \alpha_2(\mathbf{x}_{k-1})] \\
 &= \alpha_1(\mathbf{x}_k) - \alpha_1(\mathbf{x}_{k-1}) + \alpha_2(\mathbf{x}_k) - \alpha_2(\mathbf{x}_{k-1}) \\
 &= \Delta(\alpha_1)_k + \Delta(\alpha_2)_k
 \end{aligned}$$

$$\begin{aligned}
 \text{Consider } \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha) &= \sum_{k=1}^n \mathbf{M}_k \Delta \alpha_k \\
 &= \sum_{k=1}^n \mathbf{M}_k [\Delta(\alpha_1)_k + \Delta(\alpha_2)_k] \\
 &= \sum_{k=1}^n \mathbf{M}_k \Delta(\alpha_1)_k + \sum_{k=1}^n \mathbf{M}_k \Delta(\alpha_2)_k \\
 &= \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha_1) + \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha_2) \\
 &\leq \overline{\inf \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha_1)} + \overline{\inf \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha_2)} \\
 &= \overline{\int_a^b \mathbf{f} d\alpha_1} + \overline{\int_a^b \mathbf{f} d\alpha_2}
 \end{aligned}$$

Therefore ,

$$\inf \mathbf{U}(\mathbf{P}, \mathbf{f}, \alpha) \leq \overline{\int_a^b \mathbf{f} d\alpha_1} + \overline{\int_a^b \mathbf{f} d\alpha_2}$$

Therefore,

$$\overline{\int_a^b \mathbf{f} d\alpha} \leq \overline{\int_a^b \mathbf{f} d\alpha_1} + \overline{\int_a^b \mathbf{f} d\alpha_2}$$

Therefore,

$$\int_a^b \mathbf{f} d\alpha \leq \int_a^b \mathbf{f} d\alpha_1 + \int_a^b \mathbf{f} d\alpha_2 \qquad \dots\dots(1)$$

Similarly considering the lower sums , we can prove that

$$\int_a^b f d\alpha \geq \int_a^b f d\alpha_1 + \int_a^b f d\alpha_2 \quad \dots\dots(2)$$

From (1) and (2),

$$\int_a^b f d\alpha = \int_a^b f d\alpha_1 + \int_a^b f d\alpha_2 .$$

Therefore,

$$\int_a^b f d(\alpha_1 + \alpha_2) = \int_a^b f d\alpha_1 + \int_a^b f d\alpha_2 .$$

Hence the theorem.

**Theorem 2.7**

If  $f \in R(\alpha)$  on  $[a, b]$  and  $c$  is a positive constant ,

then  $f \in R(c\alpha)$  on  $[a, b]$  and  $\int_a^b f d(c\alpha) = c \int_a^b f d\alpha$

**Proof**

This follows from theorem 2.6

**Theorem 2.8**

If  $f \in R(\alpha)$  on  $[a, b]$  and  $g \in R(\alpha)$  on  $[a, b]$ , then

- (a)  $f^2 \in R(\alpha)$  on  $[a, b]$
- (b)  $fg \in R(\alpha)$  on  $[a, b]$
- (c)  $|f| \in R(\alpha)$  on  $[a, b]$  and  $|\int_a^b f d\alpha| \leq \int_a^b |f| d\alpha$

**Proof:**

- (a) Let  $P$  be any partition of  $[a, b]$ .  
Let  $M_k(f)$  denotes  $\sup\{ f(x)/ x \in [x_{k-1}, x_k]\}$  and  
 $m_k(f)$  denotes  $\inf\{ f(x)/ x \in [x_{k-1}, x_k]\}$

Then  $M_k(f^2) = \sup\{ f^2(x)/ x \in [x_{k-1}, x_k]\}$   
 $= [M_k(|f|)]^2$   
 $m_k(f^2) = [m_k(|f|)]^2$

Since  $f \in R(\alpha)$  on  $[a, b]$ ,

$f$  is bounded on  $[a, b]$ .

Therefore there exists a  $M > 0$  such that

$$|f(x)| \leq M.$$

Then  $M_k(f) \leq M$  and  
 $m_k(f) \leq M$ , for all  $k=1,2,\dots,n$ .

Also for given  $\epsilon > 0$ , there exists a partition  $P$  of  $[a, b]$  such that

$$U(P, f, \alpha) - L(P, f, \alpha) < \epsilon$$

Therefore ,

$$\begin{aligned} M_k(f^2) - m_k(f^2) &= [M_k(|f|)]^2 - [m_k(|f|)]^2 \\ &= [M_k(|f|) + m_k(|f|)][M_k(|f|) - m_k(|f|)] \\ &\leq 2M[M_k(|f|) - m_k(|f|)] \end{aligned}$$

Therefore ,

$$\begin{aligned} U(P, f^2, \alpha) - L(P, f^2, \alpha) &= \sum_{k=1}^n [M_k(f^2) - m_k(f^2)] \Delta\alpha_k \\ &\leq 2M \sum_{k=1}^n [M_k(f) - m_k(f)] \Delta\alpha_k \end{aligned}$$

$$\begin{aligned} &\leq 2M[U(P, f, \alpha) - L(P, f, \alpha)] \\ &< 2M(\epsilon / 2M) \\ &< \epsilon. \end{aligned}$$

Therefore,  $f^2 \in R(\alpha)$  on  $[a, b]$ .  
Hence part (a) is proved.

(b) Since  $f \in R(\alpha)$  on  $[a, b]$  and  $g \in R(\alpha)$  on  $[a, b]$ ,  
by theorems 2.1 and 2.2,  
 $f + g \in R(\alpha)$  on  $[a, b]$ , and  $f - g \in R(\alpha)$  on  $[a, b]$ .

Therefore by part (a) proved above,  
 $(f + g)^2 \in R(\alpha)$  on  $[a, b]$  and  
 $(f - g)^2 \in R(\alpha)$  on  $[a, b]$ .

Therefore again by theorems theorem 2.1 and 2.2,  
 $(1/4)[(f + g)^2 - (f - g)^2] \in R(\alpha)$  on  $[a, b]$ .  
i.e  $fg \in R(\alpha)$  on  $[a, b]$ .

Hence part (b) is proved.

(c) Since  $f \in R(\alpha)$  on  $[a, b]$ ,  
for given  $\epsilon > 0$ , there exists a partition  $P$  of  $[a, b]$  such that  
 $U(P, f, \alpha) - L(P, f, \alpha) < \epsilon$

Let  $P$  be any partition of  $[a, b]$ .

Since  $||f(x)| - |f(y)|| \leq |f(x) - f(y)|$ ,

$$\begin{aligned} M_k(|f|) - m_k(|f|) &= \sup\{ ||f(x)| - |f(y)|| / x, y \in [x_{k-1}, x_k] \} \\ &\leq \sup\{ |f(x) - f(y)| / x, y \in [x_{k-1}, x_k] \} \\ &\leq M_k(f) - m_k(f) \end{aligned}$$

Therefore,  
 $U(P, |f|, \alpha) - L(P, |f|, \alpha) \leq U(P, f, \alpha) - L(P, f, \alpha) < \epsilon$ .

Hence  $|f| \in R(\alpha)$  on  $[a, b]$ .

Now for all  $x$ ,  $f(x) \leq |f(x)|$  and  
 $-f(x) \leq |f(x)|$ .

Therefore by applying theorem 2.3, we get

$$\begin{aligned} \int_a^b f d\alpha &\leq \int_a^b |f| d\alpha \quad \text{and} \\ -\int_a^b f d\alpha &\leq \int_a^b |f| d\alpha \end{aligned}$$

Therefore  $|\int_a^b f d\alpha| \leq \int_a^b |f| d\alpha$ .

Hence part (c) is proved.  
Hence the theorem.

**Theorem 2.8**

Suppose  $\alpha$  is a strictly increasing continuous function that maps an interval  $[A, B]$  onto  $[a, b]$ .  
Suppose  $\alpha$  is monotonically increasing on  $[a, b]$  and  $f \in R(\alpha)$  on  $[a, b]$ .  
Define  $F$  and  $g$  on  $[A, B]$  by

$$F(y) = \int_a^y f(\alpha(t)) d\alpha(t), \quad g(y) = f(\alpha(y)).$$

Then  $g \in R(\alpha)$  and  $\int_A^B g d\beta = \int_a^b f d\alpha$ .

**Proof:**

To each partition  $P = \{x_0, x_1, \dots, x_n\}$  of  $[a, b]$ , there exists a partition  $Q = \{y_0, y_1, \dots, y_n\}$  of  $[A, B]$  such that

$$x_i = \phi(y_i).$$

All partitions of  $[A, B]$  can be obtained in this way.

Since  $g(y) = f(\phi(y))$  on  $[A, B]$ ,

the values taken by  $g$  on  $[y_{i-1}, y_i]$  are the same as those taken by  $f$  on  $[x_{i-1}, x_i]$ .

Therefore,

$$\begin{aligned} U(Q, g, \alpha) &= U(P, f, \alpha) \text{ and} \\ L(Q, g, \alpha) &= L(P, f, \alpha). \end{aligned} \quad \dots(1)$$

Since  $f \in R(\alpha)$  on  $[a, b]$ ,

for given  $\epsilon > 0$ , there exists a partition  $P$  of  $[a, b]$  such that

$$U(P, f, \alpha) - L(P, f, \alpha) < \epsilon.$$

Therefore,

$$U(Q, g, \alpha) - L(Q, g, \alpha) = U(P, f, \alpha) - L(P, f, \alpha) < \epsilon$$

Therefore  $g \in R(\alpha)$  on  $[A, B]$ .

Moreover from (1),

$$\inf U(Q, g, \alpha) = \inf U(P, f, \alpha)$$

Therefore,

$$\overline{\int_A^B g d\beta} = \overline{\int_a^b f d\alpha}.$$

since  $f \in R(\alpha)$  on  $[a, b]$  and  $g \in R(\alpha)$  on  $[A, B]$ ,

$$\begin{aligned} \int_A^B g d\beta &= \overline{\int_A^B g d\beta} \text{ and} \\ \int_a^b f d\alpha &= \overline{\int_a^b f d\alpha}. \end{aligned}$$

Therefore,

$$\int_A^B g d\beta = \int_a^b f d\alpha.$$

Hence the theorem.

### Check your progress

1. If  $a < b$ ,  $\int_b^a f d\alpha = ?$
2. Suppose  $f$  is a bounded valued function on  $[a, b]$  and  $f^2 \in R$  on  $[a, b]$ . Does it follow that  $f \in R$  on  $[a, b]$ .

### 2.3 Let us sum up

In this lesson, we have seen that

- Sum of two Riemann Stieltjes integrable functions is also Riemann Stieltjes integrable.
- Scalar product of a Riemann Stieltjes integrable function is also Riemann Stieltjes integrable.
- Modulus of a Riemann Stieltjes integrable function is also Riemann Stieltjes integrable.
- Square of a Riemann Stieltjes integrable function is also Riemann Stieltjes integrable and

- If a function is Riemann Stieltjes integrable on an interval, then it is also Riemann Stieltjes integrable on any of its subinterval.

#### 2.4 Lesson End Activities

1. Show that  $\int_0^1 x^2 dx = 3/5$  where  $\alpha(x) = x^3$
2. Show that  $\int_0^2 [x] dx = 3/5$  where  $\alpha(x) = x^2 = 3$ .

#### 2.5 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## LESSON – 3

### INTEGRATION AND DIFFERENTIATION

#### 3.0 Introduction

#### 3.1 Aims and objectives

#### 3.2 Integration and differentiation

#### 3.3 Let us sum up

#### 3.4 Lesson End Activities

#### 3.5 References

#### 3.0 Introduction

In this lesson, we are going to see that differentiation and integration, are in a certain sense, inverse operations.

#### 3.1 Aims and objectives

After studying this lesson, you would know

- the fundamental theorems on calculus and
- the theorem on integration by parts.

#### 3.2 Integration and differentiation

### Theorem 3.1

Let  $f \in R$  on  $[a, b]$  (i.e  $f$  is Riemann-integrable on  $[a, b]$ ).

For  $a \leq x \leq b$ , define  $F(x) = \int_a^x f(t) dt$ .

Then  $F$  is continuous on  $[a, b]$ ;

Furthermore, if  $f$  is continuous at a point  $x_0$  of  $[a, b]$ , then  $F$  is differentiable at  $x_0$  and

$$F'(x_0) = f(x_0).$$

#### Proof:

Since  $f \in R$  on  $[a, b]$ ,

$f$  is bounded.

Therefore there exists  $M > 0$  such that

$$|f(t)| \leq M \quad \text{for } a \leq t \leq b.$$

Therefore if  $a \leq x \leq y \leq b$ , then

$$\begin{aligned} |F(y) - F(x)| &= \left| \int_a^y f(t) dt - \int_a^x f(t) dt \right| \\ &= \left| \int_x^y f(t) dt \right| \\ &\leq \int_x^y |f(t)| dt \\ &\leq M \int_x^y dt = M(y-x) \end{aligned}$$

i.e  $|F(y) - F(x)| \leq M(y-x)$ .

Therefore,

$$|F(y) - F(x)| < \varepsilon \quad \text{provided that } |y-x| < \varepsilon/M.$$

Hence  $F$  is continuous on  $[a, b]$ .

Suppose if  $f$  is continuous at  $x_0$ .

then for given  $\varepsilon > 0$ , there exists a  $\delta > 0$  such that

$$|f(t) - f(x_0)| < \varepsilon \quad \text{whenever } |t - x_0| < \delta.$$

Hence if  $x_0 - \delta < s \leq x_0 \leq t < x_0 + \delta$  and  $a \leq s < t \leq b$ ,

$$\begin{aligned} \left| \frac{F(t) - F(s)}{t - s} - f(x_0) \right| &= \left| \frac{1}{t - s} \int_s^t f(t) dt - f(x_0) \right| \\ &= \left| \frac{1}{t - s} \int_s^t [f(t) - f(x_0)] dt \right| \\ &\leq \frac{1}{t - s} \int_s^t |f(t) - f(x_0)| dt \\ &< \frac{\varepsilon}{t - s} \int_s^t dt \\ &< \frac{\varepsilon}{t - s} [t - s] \\ &< \varepsilon \end{aligned}$$

Therefore ,

$$\left| \frac{F(t) - F(s)}{t - s} - f(x_0) \right| < \varepsilon \quad \text{whenever } x_0 - \delta < s \leq x_0 \leq t < x_0 + \delta$$

Therefore  $F'(x_0) = f(x_0)$ .

Hence the theorem.

### Theorem 3.2 Fundamental theorem of calculus.

If  $f \in R$  on  $[a, b]$  and if there exists a differentiable function  $F$  on  $[a, b]$  such that  $F' = f$ , then

$$\int_a^b f(x) dx = F(b) - F(a).$$

**Proof:**

Let  $\varepsilon > 0$  be given.

Since  $f \in R$  on  $[a, b]$ ,

there exists a partition  $P = \{x_0, x_1, x_2, \dots, x_n\}$  of  $[a, b]$  such that

$$U(P, f) - L(P, f) < \varepsilon.$$

By the Mean value theorem , there exist points  $t_i$  such that

$$\begin{aligned} F(x_i) - F(x_{i-1}) &= f(t_i)[x_i - x_{i-1}] \\ &= f(t_i) \Delta x_i \end{aligned} \quad \text{for } i=1, 2, \dots, n.$$

Therefore,

$$\begin{aligned} \sum_{i=1}^n f(t_i) \Delta x_i &= \sum_{i=1}^n [F(x_i) - F(x_{i-1})] \\ &= F(x_n) - F(x_0) \\ &= F(b) - F(a). \end{aligned}$$

Also 
$$L(P, f) \leq \sum_{i=1}^n f(t_i) \Delta x_i \leq U(P, f)$$

$$\text{And} \quad \mathbf{L(P,f)} \leq \int_a^b \mathbf{f(x)dx} \leq \mathbf{U(P,f)}$$

Therefore,

$$\begin{aligned} \left| \int_a^b \mathbf{f(x)dx} - [\mathbf{F(b)-F(a)}] \right| &= \left| \int_a^b \mathbf{f(x)dx} - \sum_{i=1}^n \mathbf{f(t_i)\Delta x_i} \right| \\ &\leq \mathbf{U(P,f)-L(P,f)} \\ &< \varepsilon. \end{aligned}$$

Since  $\varepsilon$  was arbitrary ,

$$\int_a^b \mathbf{f(x)dx} = \mathbf{F(b)-F(a)}.$$

Hence the theorem.

### Theorem3.3 Integration by parts.

Suppose F and G are differentiable functions on [a, b],  $F' = f \in \mathbf{R}$  and  $G' = g \in \mathbf{R}$ .

$$\text{Then} \quad \int_a^b \mathbf{F(x)g(x)dx} = \mathbf{F(b)G(b) - F(a)G(a)} - \int_a^b \mathbf{f(x)G(x)dx}.$$

**Proof:**

$$\text{Let } \mathbf{H(x) = F(x)G(x)}.$$

$$\begin{aligned} \text{Then} \quad \mathbf{H'} &= \mathbf{(FG)'} \\ &= \mathbf{FG' + F'G} \\ &= \mathbf{Fg + fG} \in \mathbf{R}. \end{aligned}$$

Applying theorem3.2 to the differentiable function H,

$$\int_a^b [\mathbf{F(x)g(x) + f(x)G(x)}] \mathbf{dx} = \mathbf{H(b)-H(a)}.$$

Therefore,

$$\int_a^b \mathbf{F(x)g(x)dx} + \int_a^b \mathbf{f(x)G(x)dx} = \mathbf{F(b)G(b)-F(a)G(a)}.$$

Therefore,

$$\int_a^b \mathbf{F(x)g(x)dx} = \mathbf{F(b)G(b) - F(a)G(a)} - \int_a^b \mathbf{f(x)G(x)dx}.$$

Hence the theorem.

### Check your progress

Derive the identities

1.  $\sum_{k=1}^n \frac{1}{k^s} = \frac{1}{n^{s-1}} + s \int_1^n \frac{[x]}{x^{s+1}} dx$ , if  $s > 1$
2.  $\sum_{k=1}^n \frac{1}{k} = \log n - \int_1^n \frac{x - [x]}{x^2} dx + 1$

### 3.3 Let us sum up

In this lesson, you have studied the important fundamental theorems on calculus that

- if F is the integral of f with respect to  $x$  on [a,b], then differentiation of F at a point gives f provided f is continuous at that point.
- If F is the integral of f with respect to  $x$  on [a,b], then  $\int_a^b \mathbf{f(x)dx} = \mathbf{F(b)-F(a)}$  and
- the result similar to integration by parts you have studied in School.

### 3.4 Lesson End Activities

If it is a real, Continously differentiable function on  $[a, b]$  with  $f(a) = f(b) = 0$  and  $\int_a^b f^2(x) dx = 1$ , prove that  $\int_a^b x f(x) f'(x) dx = -\frac{1}{2}$ .

### 3.5 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## LESSON – 4

### INTEGRATION OF VECTOR VALUED FUNCTIONS.

#### 4.0 Introduction

#### 4.1 Aim and objectives

#### 4.2 Integration of vector valued functions.

#### 4.3 Let us sum up

#### 4.4 Lesson End Activities

#### 4.5 References

#### 4.0 Introduction

In this lesson , we are going to see how to integrate vector valued functions and whether the theorems on integration we have already seen are valid for these functions.

#### 4.1 Aim and objectives

After studying this lesson you would know

- The definition of integration of vector valued functions
- Sum of the integrals of two vector valued functions is the sum of the integrals
- Integral of a scalar multiple of a vector valued function is the scalar multiple of the integral and
- Most of the results that are true for real functions is true for vector valued functions also.

#### 4.2 Integration of vector valued functions.

##### Definition:

Let  $f_1, f_2, \dots, f_k$  be real valued functions on  $[a, b]$ .

Let  $\mathbf{f} = (f_1, f_2, \dots, f_k)$  be the corresponding mapping of  $[a, b]$  onto  $\mathbb{R}^k$ .

If  $\alpha$  increases monotonically on  $[a, b]$  and if  $f_j \in \mathbf{R}(\alpha)$  for  $j=1, 2, \dots, k$ ,

then we say that  $\mathbf{f} \in \mathbf{R}(\alpha)$  on  $[a, b]$  and define  $\int_a^b \mathbf{f} d\alpha$  as

$$\int_a^b \mathbf{f} d\alpha = \left( \int_a^b f_1 d\alpha, \int_a^b f_2 d\alpha, \dots, \int_a^b f_k d\alpha \right)$$

i.e  $\int_a^b \mathbf{f} d\alpha$  is the point in  $\mathbb{R}^k$  whose  $j$ th coordinate is  $\int_a^b f_j d\alpha$ .

Let  $\mathbf{f} = (f_1, f_2, \dots, f_k)$  and  $\mathbf{g} = (g_1, g_2, \dots, g_k)$  be vector valued functions on  $[a, b]$ .

Then by the way in which we have defined  $\int_a^b \mathbf{f} d\alpha$  and using the results we have seen in Lesson 2 and 3, we get the following results.

##### Theorem 4.1

If  $\mathbf{f} \in \mathbf{R}(\alpha)$  and  $\mathbf{g} \in \mathbf{R}(\alpha)$  on  $[a, b]$ , then  $\mathbf{f} + \mathbf{g} \in \mathbf{R}(\alpha)$  and

$$\int_a^b (\mathbf{f} + \mathbf{g}) d\alpha = \int_a^b \mathbf{f} d\alpha + \int_a^b \mathbf{g} d\alpha,$$

##### Proof:

Since  $\mathbf{f} \in \mathbf{R}(\alpha)$  and  $\mathbf{g} \in \mathbf{R}(\alpha)$  on  $[a, b]$ ,

Hence  $f_j, g_j \in \mathbf{R}(\alpha)$  on  $[a, b]$ , for  $j=1,2,\dots,k$ .  
 Hence  $f_j+g_j \in \mathbf{R}(\alpha)$  on  $[a, b]$ , for  $j=1,2,\dots,k$ .  
 Therefore,  
 $(\mathbf{f}+\mathbf{g}) \in \mathbf{R}(\alpha)$  on  $[a, b]$

$$\begin{aligned} \text{And } \int_a^b (f + g)d\alpha &= (\int_a^b (f_1 + g_1)d\alpha, \int_a^b (f_2 + g_2)d\alpha, \dots, \int_a^b (f_k + g_k)d\alpha) \\ &= (\int_a^b f_1d\alpha + \int_a^b g_1d\alpha, \int_a^b f_2d\alpha + \int_a^b g_2d\alpha, \dots, \int_a^b f_kd\alpha + \int_a^b g_kd\alpha) \\ &= (\int_a^b f_1d\alpha, \int_a^b f_2d\alpha, \dots, \int_a^b f_kd\alpha) + (\int_a^b g_1d\alpha, \int_a^b g_2d\alpha, \dots, \int_a^b g_kd\alpha) \\ &= \int_a^b fd\alpha + \int_a^b gd\alpha \end{aligned}$$

Hence the theorem.

In the similar way we can prove the following results.

**Theorem4.2**

If  $\mathbf{f} \in \mathbf{R}(\alpha)$  on  $[a, b]$ , then  $c\mathbf{f} \in \mathbf{R}(\alpha)$  on  $[a, b]$ , for any constant  $c$  and

$$\int_a^b c\mathbf{f}d\alpha = c \int_a^b \mathbf{f}d\alpha .$$

**Theorem4.3**

If  $\mathbf{f} \in \mathbf{R}(\alpha)$  on  $[a, b]$  and if  $a < c < b$ , then  $\mathbf{f} \in \mathbf{R}(\alpha)$  on  $[a, c]$  and on  $[c, b]$ , and

$$\int_a^c \mathbf{f}d\alpha + \int_c^b \mathbf{f}d\alpha = \int_a^b \mathbf{f}d\alpha .$$

**Theorem4.4**

If  $\mathbf{f} \in \mathbf{R}(\alpha_1)$  on  $[a, b]$  and  $\mathbf{f} \in \mathbf{R}(\alpha_2)$  on  $[a, b]$ , then  $\mathbf{f} \in \mathbf{R}(\alpha_1 + \alpha_2)$  on  $[a, b]$  and

$$\int_a^b \mathbf{f}d(\alpha_1 + \alpha_2) = \int_a^b \mathbf{f}d\alpha_1 + \int_a^b \mathbf{f}d\alpha_2$$

**Theorem4.5**

If  $f \in R(\alpha)$  on  $[a, b]$  and  $c$  is a positive constant, then  $f \in R(c\alpha)$  on  $[a, b]$  and

$$\int_a^b fd(c\alpha) = c \int_a^b fd\alpha$$

**Theorem4.6**

Let  $f \in R$  on  $[a, b]$  (i.e  $f$  is Riemann-integrable on  $[a, b]$ ).

For  $a \leq x \leq b$ , define  $F(x) = \int_a^x \mathbf{f}(t)dt$ .

Then  $F$  is continuous on  $[a, b]$ ;

Furthermore, if  $f$  is continuous at a point  $x_0$  of  $[a, b]$ , then  $F$  is differentiable at  $x_0$  and  $F'(x_0) = f(x_0)$ .

**Theorem4.7**

If  $\mathbf{f}$  and  $\mathbf{F}$  maps  $[a, b]$  onto  $\mathbf{R}^k$ , if  $\mathbf{f} \in \mathbf{R}$  on  $[a, b]$  and if  $\mathbf{F}' = \mathbf{f}$ , then

$$\int_a^b \mathbf{f}(t)dt = \mathbf{F}(b) - \mathbf{F}(a).$$

**Theorem4.8**

If  $\mathbf{f}$  maps  $[a, b]$  onto  $\mathbb{R}^k$ , and  
 if  $\mathbf{f} \in \mathbf{R}(\alpha)$  for some monotonically increasing function  $\alpha$  on  $[a, b]$ ,  
 Then  $|\mathbf{f}| \in \mathbf{R}(\alpha)$  on  $[a, b]$  and

$$\left| \int_a^b \mathbf{f} d\alpha \right| \leq \int_a^b |\mathbf{f}| d\alpha .$$

**Proof**

If  $f_1, f_2, \dots, f_k$  are the components of  $\mathbf{f}$ , then

$$|\mathbf{f}| = (f_1^2 + f_2^2 + \dots + f_k^2)^{1/2} .$$

Since  $\mathbf{f} \in \mathbf{R}(\alpha)$  on  $[a, b]$ ,

By definition,

$$\int_a^b f_j d\alpha \in \mathbb{R} \text{ for } j=1, 2, \dots, k \text{ and}$$

By theorem 2.8,

$$\int_a^b f_j^2 d\alpha \in \mathbb{R} \text{ for } j=1, 2, \dots, k .$$

Hence by theorem 2.1,

$$\int_a^b (f_1^2 + f_2^2 + \dots + f_k^2) d\alpha \in \mathbb{R} .$$

Hence

$$|\mathbf{f}| = (f_1^2 + f_2^2 + \dots + f_k^2)^{1/2} \in \mathbf{R}(\alpha) .$$

[Since square root of a continuous function is continuous on  $[0, M]$ , for every real  $M$ ]

To prove that  $\left| \int_a^b \mathbf{f} d\alpha \right| \leq \int_a^b |\mathbf{f}| d\alpha$ .

Let  $\mathbf{y} = (y_1, y_2, \dots, y_k)$ , where  $y_j = \int_a^b f_j d\alpha$

Then  $\mathbf{y} = \int_a^b \mathbf{f} d\alpha$ .

And 
$$\begin{aligned} |\mathbf{y}|^2 &= \sum_{j=1}^k y_j^2 \\ &= \sum_{j=1}^k y_j \int_a^b f_j d\alpha \\ &= \int_a^b \left( \sum_{j=1}^k y_j f_j \right) d\alpha \end{aligned}$$

From Schwarz inequality,

$$\sum_{j=1}^k y_j f_j \leq |\mathbf{y}| |\mathbf{f}(t)| \quad (a \leq t \leq b)$$

Therefore

$$|\mathbf{y}|^2 \leq |\mathbf{y}| \int_a^b |\mathbf{f}| d\alpha$$

Therefore if  $\mathbf{y} \neq 0$ , dividing this inequality by  $|\mathbf{y}|$ , we get

$$|\mathbf{y}| \leq \int_a^b |\mathbf{f}| d\alpha .$$

Hence

$$\left| \int_a^b \mathbf{f} d\alpha \right| \leq \int_a^b |\mathbf{f}| d\alpha .$$

Hence the theorem.

**4.3 Let us sum up**

From what we have seen, it is clear that most of the properties of integrals of real or complex valued functions are valid for vector valued functions also.

#### 4.4 Lesson End Activity

if  $f : [0, 1] \rightarrow \mathbb{R}^3$  is defined by  $f(t) = (t, t^2, t^3)$  & if  $\omega(t) = t \forall t \in [0, 1]$ , find  $\int_0^1 f \cdot \omega \, dx$

#### 4.5 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Wiley and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## Lesson - 5

### RECTIFIABLE CURVES

#### 5.0 Introduction

#### 5.1 Aims and objectives

#### 5.2 Rectifiable curves

#### 5.3 Let us sum up

#### 5.4 Lesson End Activities

#### 5.5 References

#### 5.0 Introduction

In this lesson you are going to study about rectifiable curves.

#### 5.1 Aims and objectives

After studying this lesson you would know

- What is a rectifiable curve
- Length of a rectifiable curve and that
- The length is given by a Riemann integral.

#### 5.2 Rectifiable curves

##### Definition:

A continuous mapping  $\gamma$  of an interval  $[a, b]$  into  $\mathbb{R}^k$  is called a curve in  $\mathbb{R}^k$ .  
or  $\gamma$  is a curve on  $[a, b]$ .

If  $\gamma$  is one-to-one,  $\gamma$  is called an arc.

If  $\gamma(a) = \gamma(b)$ ,  $\gamma$  is called a closed curve.

##### Definition:

To each partition  $P = \{x_0, x_1, x_2, \dots, x_n\}$  of  $[a, b]$  and to each curve  $\gamma$  on  $[a, b]$ ,

we associate a number  $\Lambda(P, \gamma) = \sum_{i=1}^n |\gamma(x_i) - \gamma(x_{i-1})|$

where  $|\gamma(x_i) - \gamma(x_{i-1})| =$  distance between the points  $\gamma(x_{i-1})$  and  $\gamma(x_i)$

$\Lambda(P, \gamma) =$  length of a polygonal path with vertices at  $\gamma(x_0), \gamma(x_1), \dots, \gamma(x_n)$ .

As the partition  $P$  becomes finer and finer,

the polygon approaches  $\gamma$  more and more closely.

The length of  $\Lambda$  is defined as

$$\Lambda(\gamma) = \sup \Lambda(P, \gamma),$$

where the supremum is taken over all partitions of  $[a, b]$ .

If  $\Lambda(\gamma) < \infty$ ,  $\gamma$  is said to be rectifiable.

### Theorem 5.1

If  $\gamma'$  is continuous on  $[a, b]$ , then  $\gamma$  is rectifiable, and  $\Lambda(\gamma) = \int_a^b |\gamma'(t)| dt$ .

**Proof:**

If  $a \leq x_{i-1} < x_i \leq b$ , then

$$\begin{aligned} |\gamma(x_i) - \gamma(x_{i-1})| &= \left| \int_{x_{i-1}}^{x_i} \gamma'(t) dt \right| \\ &\leq \int_{x_{i-1}}^{x_i} |\gamma'(t)| dt \end{aligned}$$

Hence 
$$\Lambda(P, \gamma) = \sum_{i=1}^n |\gamma(x_i) - \gamma(x_{i-1})|$$

$$\leq \sum_{i=1}^n \int_{x_{i-1}}^{x_i} |\gamma'(t)| dt$$

$$\leq \int_a^b |\gamma'(t)| dt$$

for every partition P of [a,b].

Therefore ,

$$\Lambda(\gamma) \leq \int_a^b |\gamma'(t)| dt \quad \dots(1)$$

To prove the opposite inequality,

let  $\varepsilon > 0$  be given.

Since  $\gamma'$  is uniformly continuous on [a, b], there exists a  $\delta > 0$  such that

$$|\gamma'(s) - \gamma'(t)| < \varepsilon \quad \text{whenever } |s - t| < \delta.$$

Let  $P = \{x_0, x_1, \dots, x_n\}$  be a partition of [a, b], with  $\Delta x_i < \delta$  for all i.

Therefore if  $x_{i-1} \leq t \leq x_i$ ,

$$|\gamma'(t) - \gamma'(x_i)| < \varepsilon$$

Therefore,

$$\begin{aligned} |\gamma'(t) - \gamma'(x_i)| &\leq |\gamma'(t) - \gamma'(x_i)| \\ &< \varepsilon \end{aligned}$$

Therefore,

$$|\gamma'(t)| \leq |\gamma'(x_i)| + \varepsilon.$$

Hence

$$\begin{aligned} \int_{x_{i-1}}^{x_i} |\gamma'(t)| dt &\leq \int_{x_{i-1}}^{x_i} (|\gamma'(x_i)| + \varepsilon) dt \\ &\leq |\gamma'(x_i)| \Delta x_i + \varepsilon \Delta x_i \\ &\leq \left| \int_{x_{i-1}}^{x_i} \gamma'(x_i) dt \right| + \varepsilon \Delta x_i \\ &\leq \left| \int_{x_{i-1}}^{x_i} [\gamma'(t) + \gamma'(x_i) - \gamma'(t)] dt \right| + \varepsilon \Delta x_i \\ &\leq \left| \int_{x_{i-1}}^{x_i} \gamma'(t) dt \right| + \left| \int_{x_{i-1}}^{x_i} [\gamma'(x_i) - \gamma'(t)] dt \right| + \varepsilon \Delta x_i \\ &\leq |\gamma(x_i) - \gamma(x_{i-1})| + \varepsilon \Delta x_i + \varepsilon \Delta x_i \\ &\leq |\gamma(x_i) - \gamma(x_{i-1})| + 2\varepsilon \Delta x_i \end{aligned}$$

Therefore ,

$$\begin{aligned} \int_a^b |\gamma'(t)| dt &= \sum_{i=1}^n \int_{x_{i-1}}^{x_i} |\gamma'(t)| dt \\ &\leq \sum_{i=1}^n |\gamma(x_i) - \gamma(x_{i-1})| + 2\varepsilon \sum_{i=1}^n \Delta x_i \end{aligned}$$

$$\begin{aligned} &\leq \Lambda(\mathbf{P}, \gamma) + 2\varepsilon(\mathbf{b} - \mathbf{a}) \\ &\leq \Lambda(\gamma) + 2\varepsilon(\mathbf{b} - \mathbf{a}). \end{aligned}$$

Since  $\varepsilon > 0$  was arbitrary,

$$\int_a^b |\gamma'(t)| dt \leq \Lambda(\gamma) \quad \dots(2).$$

From (1) and (2), we get

$$\Lambda(\gamma) = \int_a^b |\gamma'(t)| dt$$

Hence the theorem

### Check your progress

Let  $\gamma_1, \gamma_2, \gamma_3$  be curves in the complex plane, defined on  $[0, 2]$  by  $\gamma_1(t) = e^{it}, \gamma_2(t) = e^{2it}, \gamma_3(t) = e^{2i \sin(1/t)}$ .

Show that

1. These three curves have the same range
2.  $\gamma_1$  and  $\gamma_2$  are rectifiable.
3. length of  $\gamma_1$  is  $2$ .
4. length of  $\gamma_2$  is  $4$  and
5.  $\gamma_3$  is not rectifiable.

### 5.3 Let us sum up

In this lesson you have studied

- the definition of a rectifiable curve,
- its length and that
- its length is equal to the Riemann integral of the modulus of the derivative of the rectifiable curve.

### 5.4 Lesson End Activities

if  $\gamma_1, \gamma_2$  are curves in the complex plane defined on  $[0, 1]$  &  $[0, 2]$  respectively by  $\gamma_1(t) = e^{2\pi i t}, \gamma_2(t) = e^{2\pi i t}, t \in [0, 1]$ , Prove that length of  $\gamma_2$  is twice that of  $\gamma_1$ .

### 5.5 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## UNIT II

### SEQUENCES AND SERIES OF FUNCTIONS

#### LESSON – 6

#### UNIFORM CONVERGENCE AND CONTINUITY

##### 6.0 Introduction

##### 6.1 Aims and objectives

##### 6.2 Pointwise convergence and Uniform convergence

##### 6.3 Uniform convergence and Continuity

##### 6.4 Let us sum up

##### 6.5 Lesson End Activities

##### 6.6 References

##### 6.0 Introduction

In this lesson, we are going to see about uniform convergence of a sequence of functions and continuity of the limit function.

##### 6.1 Aims and objectives

After studying this unit, you should be able to identify

- pointwise convergence of a sequence of functions
- uniform convergence of a sequence of functions
- the limit function
- continuity of the limit function

##### 6.2 Pointwise convergence and Uniform convergence

##### Definition: pointwise convergence of a sequence of functions

Suppose  $\{f_n\}$ ,  $n=1,2,\dots$  is a sequence of functions defined on a set  $E$ , and suppose that the sequence of numbers  $\{f_n(x)\}$  converges for each  $x$  in  $E$ .

Define the function  $f$  by

$$f(x) = \lim_{n \rightarrow \infty} f_n(x) \quad \text{for each } x \in E$$

Then we say that  $\{f_n\}$  converges pointwise to  $f$  on  $E$  and  $f$  is called the limit function.

##### Definition: pointwise convergence of a series of functions

Suppose  $\{f_n\}$ ,  $n=1,2,\dots$  is a sequence of functions defined on a set  $E$ , and suppose

that the series  $\sum_{n=1}^{\infty} f_n(x)$  converges for each  $x$  in  $E$

$$\text{and } f(x) = \sum_{n=1}^{\infty} f_n(x), \quad \text{for each } x \in E,$$

Then  $f$  is called the sum of the series  $\sum f_n$ .

##### Examples

1. Let  $f_n(x) = x^n$ ,  $0 < x < 1$ , for  $n=1,2,3,\dots$

Then  $\lim_{n \rightarrow \infty} f_n(x) = \lim_{n \rightarrow \infty} x^n = 0$ , if  $0 < x < 1$

$$=1, \quad \text{if } x=1.$$

Hence the limit function  $f$  is

$$f(x)=0, \quad \text{if } 0 < x < 1 \\ =1, \quad \text{if } x=1.$$

$$2. \text{ Let } f_n(x) = \frac{x^2}{(1+x^2)^n} \quad x \text{ real, } n=1,2,3,\dots$$

$$\begin{aligned} \text{Then } \sum_{n=0}^{\infty} f_n(x) &= \sum_{n=0}^{\infty} \frac{x^2}{(1+x^2)^n} \\ &= x^2 \left[ 1 - \frac{1}{1+x^2} \right]^{-1} \\ &= x^2 \left[ \frac{1+x^2-1}{1+x^2} \right]^{-1} \\ &= 1+x^2 \quad \text{if } x \neq 0. \end{aligned}$$

$$\text{If } x=0, \quad \sum_{n=0}^{\infty} f_n(0) = 0$$

Hence sum of the series is

$$f(x)=0 \quad \text{if } x=0 \\ =1 \quad \text{if } x \neq 0$$

### Definition: Uniform convergence of a sequence of functions

A sequence of functions  $\{f_n\}$ ,  $n=1,2,3,\dots$  is said to converge uniformly on  $E$  to a function  $f$ , if for every  $\epsilon > 0$ , there exists an integer  $N$  such that  $n > N$  implies

$$|f_n(x) - f(x)| < \epsilon \quad \text{for all } x \in E.$$

we write this as  $f_n \rightarrow f$  uniformly on  $E$ .

Remark:

Here  $N$  depends only on  $\epsilon$  and not on  $x$ .

Hence the convergence is said to be uniform.

Otherwise the convergence is only pointwise convergence.

### Theorem 6.2.1 Cauchy criterion for uniform convergence

The sequence of functions  $\{f_n\}$ , defined on  $E$  converges uniformly on  $E$  if and only if there exists an integer  $N$  such that  $m > N$ ,  $n > N$ ,  $x \in E$  implies

$$|f_n(x) - f_m(x)| < \epsilon.$$

#### Proof

Suppose that  $\{f_n\}$  converges uniformly on  $E$  to a limit function  $f$ .

Then by definition of uniform convergence,

For given  $\epsilon > 0$ , there exists an  $N$  such that  $n > N$  implies

$$|f_n(x) - f(x)| < \epsilon/2 \quad \text{for all } x \in E \quad \dots(1)$$

Therefore if  $m > N$ ,  $n > N$ ,  $x \in E$  then

$$\begin{aligned} |f_n(x) - f_m(x)| &= |(f_n(x) - f(x)) - (f_m(x) - f(x))| \\ &= |f_n(x) - f(x)| + |f_m(x) - f(x)| \\ &< \epsilon/2 + \epsilon/2 \end{aligned}$$

Conversely,

assume that for given  $\epsilon > 0$ , there exists an integer  $N$  such that that  $m > N, n > N, x \in E$  implies

$$|f_n(x) - f_m(x)| < \epsilon \quad \dots(2)$$

Then for every  $x$  in  $E$ ,

$\{f_n(x)\}$  is a Cauchy sequence of numbers.

Since a Cauchy sequence of real or complex numbers converges,

$\{f_n(x)\}$  converges for every  $x$  in  $E$ .

Define  $f$  on  $E$  as

$$f(x) = \lim_{n \rightarrow \infty} f_n(x) \text{ for each } x \text{ in } E.$$

Then  $\{f_n\}$  converges pointwise on  $E$ .

To show that the convergence is uniform.

In (2), let  $n$  be fixed and let  $m \rightarrow \infty$ ,

Then  $|f_n(x) - f(x)| < \epsilon$  for all  $x$  in  $E$  and for all  $n > N$ .

Hence the theorem.

### Theorem 6.2.2

Suppose  $\lim_{n \rightarrow \infty} f_n(x) = f(x)$ .

Put  $M_n = \sup\{|f_n(x) - f(x)|, x \in E\}$

Then  $f_n \rightarrow f$  uniformly on  $E$  if and only if  $M_n \rightarrow 0$  as  $n \rightarrow \infty$ .

#### Proof

By definition  $f_n \rightarrow f$  uniformly on  $E$

if and only if,

for given  $\epsilon > 0$ , there exists  $N$  such that  $n > N$  implies  $|f_n(x) - f(x)| < \epsilon$  for all  $x \in E$ .

if and only if

$$M_n = \sup\{|f_n(x) - f(x)|, x \in E\}$$

if and only if

$$\lim_{n \rightarrow \infty} M_n = 0.$$

Hence the theorem.

### Theorem 6.2.3

Suppose  $\{f_n\}$  is a sequence of functions defined on  $E$ , and suppose that  $|f_n(x)| \leq M_n$  ( $x \in E, n=1,2,3,\dots$ ).

Then  $\sum f_n$  converges uniformly on  $E$  if  $\sum M_n$  converges.

#### Proof

$$\text{Let } t_n = M_1 + M_2 + \dots + M_n$$

$$\text{And } s_n(x) = \sum_{k=1}^n f_k(x)$$

Assume that  $\sum M_n$  converges.

Hence  $\{t_n\}$  is a Cauchy sequence.

Hence, for given  $\epsilon > 0$ , there exists  $N$  such that  $m > N, n > N$  implies

$$|t_m - t_n| < \epsilon \quad \dots(1)$$

Let  $m > n > N$

Then (1) implies that

$$\sum_{k=n+1}^m M_k < \epsilon \quad \dots(2)$$

Therefore for  $m > n > N$ ,

$$|s_m(x) - s_n(x)| = \left| \sum_{k=n+1}^m f_k(x) \right|$$

$$\leq \sum_{k=1}^n |f_k(x)|$$

$$\leq \sum_{k=n+1}^m M_k$$

for all x in E.

Hence  $\{s_n\}$  satisfies the Cauchy criterion for uniform convergence.  
 Hence  $\{s_n\}$  converges uniformly on E.  
 Hence  $\sum f_n$  converges uniformly on E.  
 Hence the theorem.

### 6.3 Uniform convergence and continuity

#### Definition: Continuity of a function

A function f defined on E is said to be continuous at  $x \in E$  if

$$\lim_{t \rightarrow x} f(t) = f(x).$$

#### Theorem 6.3.1

Suppose  $f_n \rightarrow f$  uniformly on a set E in a metric space.

Let x be a limit point of E, and suppose that

$$\lim_{t \rightarrow x} f_n(t) = A_n \quad (n=1,2,3,\dots) \quad \dots(1)$$

Then  $\{A_n\}$  converges and  $\lim_{t \rightarrow x} f(t) = \lim_{n \rightarrow \infty} A_n$ . ... (2)

In other words,  $\lim_{t \rightarrow x} \lim_{n \rightarrow \infty} f_n(t) = \lim_{n \rightarrow \infty} \lim_{t \rightarrow x} f_n(t)$  ... (3)

#### Proof

Since  $f_n \rightarrow f$  uniformly on E, by definition,  
 for given  $\epsilon > 0$ ,

there exists an integer  $N_1$  such that  $n \geq N_1$  implies

$$|f_n(t) - f(t)| < \epsilon/3 \quad \text{for all } t \in E \quad \dots(4)$$

By Cauchy's criterion for uniform convergence,  
 for the same  $\epsilon > 0$ ,

there exists an integer  $N_2$  such that  $m \geq N_2, n \geq N_2, t \in E$  implies that

$$|f_n(t) - f_m(t)| < \epsilon/3 \quad \text{for all } t \in E \quad \dots(5)$$

Letting  $t \rightarrow x$  in (5), we get

$$|A_n - A_m| < \epsilon/3 \quad \text{for } m \geq N_2, n \geq N_2 \quad \dots(6)$$

Therefore  $\{A_n\}$  is a Cauchy sequence.

Hence  $\{A_n\}$  converges, say to A.

Therefore for the same  $\epsilon > 0$  taken above,

there exists an integer  $N_3$  such that  $n \geq N_3$  implies

$$|A_n - A| < \epsilon/3 \quad \dots(7)$$

Let  $N = \max(N_1, N_2, N_3)$

Then from (4),

$$|f_N(t) - f(t)| < \epsilon/3 \quad \dots(8)$$

From (7),  $|A_N - A| < \epsilon/3$  ... (9)

By assumption (1),

$$\lim_{t \rightarrow x} f_N(t) = A_N.$$

Therefore there exists a neighborhood V of x such that

$$|f_N(t) - A_N| < \epsilon/3 \quad \text{for all } t \in V \cap E, t \neq x \quad \dots(10)$$

From (8), (9) and (10),

$$|f(t)-A|=|f(t) - f_N(t) + f_N(t) - A_N + A_N - A |$$

$$| f(t) - f_N(t)| + |f_N(t) - A_N| + |A_N - A |$$

$$/3+ /3+ /3$$

for all  $t \in V E$  and  $t \ x$ ,

Hence  $\lim_{t \rightarrow x} f(t)=A$ .  
 i.e  $\lim_{t \rightarrow x} f(t)= \lim_{n \rightarrow \infty} A_n$ .  
 i.e  $\lim_{t \rightarrow x} \lim_{n \rightarrow \infty} f_n(t)=\lim_{n \rightarrow \infty} \lim_{t \rightarrow x} f_n(t)$   
 Hence the theorem.

Remark:

The above theorem says that if  $f_n \rightarrow f$  uniformly and  $x$  is a limit point of  $E$ , then the limit processes can be interchanged.

**Theorem6.3.2**

If  $\{f_n\}$  is a sequence of continuous functions on  $E$  and if  $f_n \rightarrow f$  uniformly on  $E$ , then  $f$  is continuous.

**Proof**

Let  $\{f_n\}$  be sequence of continuous functions on  $E$ .  
 Let  $x$  be any point of  $E$ .  
 Then each  $f_n$  is continuous at  $x$ .  
 Therefore,

$$\lim_{t \rightarrow x} f_n(t)=f_n(x), \quad n=1,2,3,\dots$$

Applying the above theorem6.3.1 and replacing  $A_n$  by  $f_n(x)$ , we get

$$\lim_{t \rightarrow x} f(t)=\lim_{n \rightarrow \infty} f_n(x).$$

Since  $f_n \rightarrow f$  on  $E$ ,

$$\lim_{n \rightarrow \infty} f_n(x)=f(x).$$

Hence  $\lim_{t \rightarrow x} f(t)= f(x)$ .

Hence the theorem.

**Theorem6.3.3**

Suppose  $K$  is compact and

- (a)  $\{f_n\}$  is a sequence of continuous functions on  $K$
- (b)  $\{f_n\}$  converges pointwise to a continuous function  $f$  on  $K$ ,
- (c)  $f_n(x) \leq f_{n+1}(x)$  for all  $x \in E, n=1,2,3,\dots$

Then  $f_n \rightarrow f$  uniformly on  $K$ .

**Proof**

$$\text{Let } g_n=f_n-f, \quad n=1,2,3,\dots$$

Since  $f_n$  and  $f$  are continuous functions on  $K$ ,

$$g_n \text{ is also continuous on } K, n=1,2,3,\dots$$

Since  $f_n \rightarrow f$  pointwise on  $K$ ,

$$g_n \rightarrow 0 \text{ pointwise on } K.$$

Since  $f_n(x) \leq f_{n+1}(x)$  for all  $x \in K, n=1,2,3,\dots$

$$f_n(x) - f(x) \leq f_{n+1}(x) - f(x) \text{ for all } x \in K, n=1,2,3,\dots$$

Hence  $g_n(x) \leq g_{n+1}(x)$ , for all  $x \in K, n=1,2,3,\dots$  ...(1)

**Claim:**  $g_n \rightarrow 0$  uniformly on  $K$ .

Let  $\epsilon > 0$  be given.

$$\text{Let } K_n = \{x \in K / g_n(x) \leq \epsilon\}$$

$$=g_n^{-1}([\epsilon, \infty)) \quad \dots(2)$$

Since each  $g_n$  is continuous and inverse image of a closed set under a continuous function is closed,

each  $K_n$  is a closed subset of  $K$ .

Since  $K$  is compact and a closed subset of a compact set is closed,

each  $K_n$  is compact.

If  $x \in K_{n+1}$ , then  $g_{n+1}(x)$

Therefore,

$$g_n(x) \leq g_{n+1}(x)$$

Hence  $x \in K_n$

Hence  $K_{n+1} \subseteq K_n$

Fix  $x \in K$ .

Since  $g_n(x) > 0$ ,

for given  $\epsilon > 0$ , there exists an integer  $N_1$  such that  $n \geq N_1$  implies

$$g_n(x) > \epsilon$$

This implies that

$$x \notin K_n \text{ for all } n \geq N_1.$$

This implies that

$K_n$  is empty.

Hence there exists an integer  $N$  such that  $K_N$  is empty.

This implies that

$$g_N(x) = 0 \text{ for all } x \in K.$$

From (1), it is clear that

$\{g_n\}$  is monotonically decreasing sequence.

Therefore,

$$g_n(x) \geq g_N(x) \text{ for all } n \geq N.$$

Hence  $0 \leq g_n(x) - g_N(x) < \epsilon$  for all  $x \in K, n \geq N$

Hence,

$$g_n > 0 \text{ uniformly on } K.$$

Hence,

$$f_n - f > 0 \text{ uniformly on } K.$$

Hence  $f_n > f$  uniformly on  $K$ .

Hence the theorem.

### Check your progress

1. If  $\{f_n\}$  and  $\{g_n\}$  converge uniformly on a set  $E$ , prove that  $\{f_n + g_n\}$  converge uniformly on  $E$ .

If in addition,  $\{f_n\}$  and  $\{g_n\}$  are bounded, prove that  $\{f_n g_n\}$  converges uniformly on  $E$ .

2. Is the converse of the theorem 6.3.2 true?
3. Is the compactness necessary in theorem 6.3.3?

### Answer:

2. No.

A sequence of continuous functions may converge to a continuous function although the convergence is not uniform.

Example:

Let  $f_n(x) = n^2 x(1-x^2)^n, \quad 0 \leq x \leq 1, n=1,2,3,\dots$

Each  $f_n$  is a continuous function on  $[0, 1]$ .

And  $\lim_{n \rightarrow \infty} f_n(x) = 0 \quad 0 < x < 1$

Since  $f_n(0) = 0$ , for all  $n$ ,

$$\lim_{n \rightarrow \infty} f_n(x) = 0 \quad \text{for } x=0$$

Hence the limit function is

$$f(x) = 0 \quad 0 \leq x \leq 1.$$

Here  $f(x)$  is also a continuous function on  $[0, 1]$ .

But  $\{f_n\}$  does not converge uniformly to  $f$  on  $[0, 1]$ .

3. Yes.

Example:

$$\text{Let } f_n(x) = \frac{1}{nx+1} \quad 0 < x < 1, n=1,2,3,\dots$$

Then  $f_n(x) > 0$  monotonically in  $(0, 1)$ ,

But the convergence is not uniform.

#### 6.4 Let us sum up

In this lesson, we have seen

- Pointwise convergence of a sequence of functions.
- Uniform convergence of a sequence of functions.
- Pointwise convergence of a series of functions.
- Uniform convergence of a series of functions.

We have also seen that

- Limit of a sequence of continuous functions converging uniformly is continuous.
- If a sequence of continuous functions is monotonic and converges pointwise to a continuous function on a compact set, then the convergence is uniform.

#### 6.5 Lesson End Activities

1. Prove that the sequence of functions  $\{f_n\}_{n=1}^{\infty}$  converges pointwise but not uniformly. Given,  $f_n(x) = n/(1+n^2x^2)$ ;  $x \in [0, 1]$ .
2. Show that the sequence  $\{f_n\}_{n=1}^{\infty}$  where  $f_n(x) = \sin nx/\sqrt{n}$  is uniformly convergent on  $[0, \pi]$ .

#### 6.6 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Wiley and Sons, New York, 1976.
2. W. Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## Lesson 7 Uniform convergence and integration

### 7.0 Introduction

#### 7.1 Aims and objectives

### 7.2 Uniform convergence and integration

#### 7.3 Let us sum up

#### 7.4 Lesson End Activities

#### 7.5 References

### 7.0 Introduction

In this lesson, we are going to learn about the integrability of the limit function of a sequence of integrable functions.

#### 7.1 Aims and objectives

After studying this lesson, you would know

- whether the limit function of a sequence of integrable functions is integrable and
- if it is so, can the operations of integration and limit can be interchanged.

### 7.2 Uniform convergence and integration

#### Theorem 7.1

Let  $f$  be a monotonically decreasing function on  $[a, b]$ .  
 Suppose  $f_n \in R(\ )$  on  $[a, b]$ , for  $n=1,2,3,..$  and  
 suppose  $f_n > f$  uniformly on  $[a, b]$ ,  
 then  $f \in R(\ )$  on  $[a, b]$  and

$$\int_a^b f d\alpha = \lim_{n \rightarrow \infty} \int_a^b f_n d\alpha$$

#### Proof

Let  $\{f_n\}$  be a sequence of real valued functions on  $[a, b]$  such that

- (1)  $f_n \in R(\ )$  on  $[a, b]$ , for  $n=1,2,3,..$  and
- (2)  $f_n > f$  uniformly on  $[a, b]$

Let  $\epsilon_n = \sup_{a \leq x \leq b} |f_n(x) - f(x)|$

Then  $\epsilon_n > 0$  as  $n \rightarrow \infty$ .

This implies that

$$\begin{aligned} & |f_n(x) - f(x)| < \epsilon_n & a \leq x \leq b \\ \text{(or)} & |f(x) - f_n(x)| < \epsilon_n & a \leq x \leq b \\ \text{(i.e)} & f(x) - \epsilon_n < f_n(x) < f(x) + \epsilon_n & a \leq x \leq b \end{aligned}$$

Therefore,

$$f_n(x) - \epsilon_n < f(x) < f_n(x) + \epsilon_n \quad a \leq x \leq b \quad \dots(1)$$

Hence ,

$$f_n(x) - \epsilon_n < f(x) < f_n(x) + \epsilon_n \quad a \leq x \leq b$$

Hence ,

$$\int_a^b (f_n - \epsilon_n) d\alpha < \int_a^b f d\alpha < \int_a^b (f_n + \epsilon_n) d\alpha \quad \dots(2)$$

Similarly from (1),

$$f(x) < f_n(x) + \epsilon_n \quad a \leq x \leq b$$

Hence,

$$\int_a^b f d\alpha < \int_a^b (f_n + \epsilon_n) d\alpha \quad \dots(3)$$

Since  $f_n \in R(\ )$  on  $[a, b]$ ,  $n=1,2,3,..$

$f_n - \epsilon_n \in R(\ )$  on  $[a, b]$  and

$f_n + \varepsilon_n \in R(\ )$  on  $[a, b]$

Hence ,

$$\int_a^b (f_n - \varepsilon_n) d\alpha = \int_a^b (f_n - \varepsilon_n) d\alpha \quad \dots(4)$$

And 
$$\int_a^b (f_n + \varepsilon_n) d\alpha = \int_a^b (f_n + \varepsilon_n) d\alpha \quad \dots(5)$$

We know that,

$$\int_a^b f d\alpha = \int_a^b f d\alpha \quad \dots(6)$$

From (2),(3),(4),(5) and (6), we get that

$$\int_a^b (f_n - \varepsilon_n) d\alpha \leq \int_a^b f d\alpha \leq \int_a^b (f_n + \varepsilon_n) d\alpha$$

Therefore,

$$\int_a^b f_n d\alpha - \varepsilon_n \int_a^b d\alpha \leq \int_a^b f d\alpha \leq \int_a^b f_n d\alpha + \varepsilon_n \int_a^b d\alpha$$

$$-\varepsilon_n [(b) - (a)] \leq \int_a^b f d\alpha - \int_a^b f_n d\alpha \leq \varepsilon_n [(b) - (a)]$$

Therefore ,

$$|\int_a^b f d\alpha - \int_a^b f_n d\alpha| \leq \varepsilon_n [(b) - (a)] > 0 \quad \text{as } \varepsilon_n > 0$$

Therefore,

$$\lim_{n \rightarrow \infty} \int_a^b f_n d\alpha = \int_a^b f d\alpha ,$$

Hence ,

$$\lim_{n \rightarrow \infty} \int_a^b f_n d\alpha = \int_a^b (\lim_{n \rightarrow \infty} f_n) d\alpha .$$

Hence the theorem.

**Theorem7.2**

If  $f_n \in R(\ )$  on  $[a, b]$ , for  $n=1,2,3,..$  and if

$$f(x) = \sum_{n=1}^{\infty} f_n(x) \quad \text{a x b,}$$

and the series converges uniformly on  $[a, b]$ , then

$$\int_a^b f d\alpha = \sum_{n=1}^{\infty} \int_a^b f_n d\alpha .$$

In other words , the series can be integrated term by term.

**Proof:**

Let  $s_n(x) = \sum_{k=1}^n f_k(x)$ ,  $n=1,2,3,..$

Since  $\sum f_n$  converges uniformly to  $f$  on  $[a, b]$ ,  
 $s_n \rightarrow f$  uniformly on  $[a, b]$ .

Therefore ,

by the above theorem 7.1,

$$\begin{aligned} \int_a^b f d\alpha &= \lim_{n \rightarrow \infty} \int_a^b s_n d\alpha \\ &= \lim_{n \rightarrow \infty} \int_a^b \left( \sum_{k=1}^n f_k \right) d\alpha \\ &= \lim_{n \rightarrow \infty} \left( \sum_{k=1}^n \int_a^b f_k d\alpha \right) \\ &= \sum_{k=1}^{\infty} \int_a^b f_k d\alpha . \end{aligned}$$

Hence the theorem.

### 7.3 Let us sum up

In this lesson, we have seen that

- If a sequence of integrable functions converges uniformly to a function, then the operations integration and limit can be interchanged.
- If a series of integrable functions converges uniformly to a limit function, then the series can be integrated term by term.

### 7.4 Lesson End Activities

1. Consider the sequence  $\{f_n\}_{n=1}^{\infty}$  where  $f_n(x) = nx(1-x^2)^n$   $0 \leq x \leq 1$ . Prove that the limit of the integrals is not equal to the integral of the limit.

### 7.5 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## Lesson 8 Uniform convergence and Differentiation

### 8.0 Introduction

#### 8.1 Aims and Objectives

#### 8.2 Uniform convergence and Differentiation

#### 8.3 Let us sum up

#### 8.4 Lesson End Activities

#### 8.5 References

### 8.0 Introduction

In this lesson, we are going to learn about the differentiability of the limit function of a sequence of functions.

#### 8.1 Aims and Objectives

After studying this lesson, you would know

- whether the limit function of a sequence of differentiable functions is differentiable
- If it is so, can the operations of differentiation and limit can be interchanged.

#### 8.2 Uniform convergence and Differentiation

Uniform convergence of a sequence of differentiable functions  $\{f_n\}$  implies nothing about the sequence  $\{f_n'\}$ .

We need stronger hypotheses for the assertion that:

$$f_n \rightarrow f \text{ to imply } f_n' \rightarrow f'$$

i.e to interchange the operations of differentiation and limit.

#### Example:

Let  $f_n(x) = (\sin nx)/n$   $x$  real,  $n=1,2,3,\dots$

Then  $f(x) = \lim_n f_n(x) = 0$  for all real  $x$ .

Then  $f'(x) = 0$

And  $f_n'(x) = \cos nx$

For  $x=0$ ,  $f_n'(0) = 1$  as  $n \rightarrow \infty$

Hence  $\{f_n'\}$  does not converge to  $f'$ .

#### Theorem 8.1

Suppose  $\{f_n\}$  is sequence of functions differentiable on  $[a, b]$  and such that  $\{f_n(x_0)\}$  converges for some point  $x_0$  in  $[a, b]$ .

If  $\{f_n'\}$  converges uniformly on  $[a, b]$ ,

then  $\{f_n\}$  converges uniformly on  $[a, b]$ , to a function  $f$ , and

$$f'(x) = \lim_n f_n'(x), \quad a \leq x \leq b.$$

#### Proof:

Since  $\{f_n(x_0)\}$  converges for some  $x_0$  in  $[a, b]$ , for given  $\epsilon > 0$ ,

there exists an integer  $N_1$  such that  $n > N_1, m > N_1$  implies

$$|f_n(x_0) - f_m(x_0)| < \epsilon/2 \quad \dots(1)$$

Since  $\{f_n'\}$  converge uniformly on  $[a, b]$ ,

$\{f_n'\}$  satisfies Cauchy criterion for uniform convergence.

Therefore for the same  $\epsilon > 0$ ,

there exists an integer  $N_2$  such that that  $n > N_2, m > N_2$  implies

$$|f_n'(t) - f_m'(t)| < \frac{\epsilon}{2(b-a)} \quad a < t < b \quad \dots(2)$$

Let  $N = \max(N_1, N_2)$

Let  $x, t \in [a, b]$ .

Applying mean value theorem to the function  $(f_n - f_m)$ , we get

$$(f_n - f_m)(x) - (f_n - f_m)(t) = (f_n - f_m)'(\xi)(x - t), \text{ where } \xi \text{ lies between } x \text{ and } t$$

Therefore,

$$|(f_n - f_m)(x) - (f_n - f_m)(t)| = |(f_n - f_m)'(\xi)(x - t)|$$

Therefore,

for any  $x, t$  in  $[a, b]$ ,  $m > N, n > N$  implies

$$\begin{aligned} |(f_n(x) - f_m(x)) - (f_n(t) - f_m(t))| &= |(f_n - f_m)'(\xi)(x - t)| \\ &\leq \frac{\epsilon}{2(b-a)} |x - t| \\ &\leq \frac{\epsilon}{2} \end{aligned} \quad \dots(3)$$

Therefore by choosing  $t = x_0$ ,

for any  $x$  in  $[a, b]$  and  $m > N, n > N$ , we get

$$\begin{aligned} |f_n(x) - f_m(x)| &= |(f_n(x) - f_m(x)) - (f_n(x_0) - f_m(x_0)) + (f_n(x_0) - f_m(x_0))| \\ &= |(f_n(x) - f_m(x)) - (f_n(x_0) - f_m(x_0))| + |(f_n(x_0) - f_m(x_0))| \\ &< \epsilon/2 + \epsilon/2 \\ &= \epsilon \end{aligned}$$

Therefore  $\{f_n\}$  satisfies Cauchy criterion for uniform convergence.

Therefore  $\{f_n\}$  converges uniformly on  $[a, b]$ .

Let  $f(x) = \lim_{n \rightarrow \infty} f_n(x) \quad a < x < b$

Fix a point  $x$  in  $[a, b]$ .

$$\text{Define } g_n(t) = \frac{f_n(t) - f_n(x)}{t - x}, \quad a < t < b, t \neq x \quad \dots(4)$$

$$\text{And } g(t) = \frac{f(t) - f(x)}{t - x}, \quad a < t < b, t \neq x \quad \dots(5)$$

Then,

$$\begin{aligned} \lim_{t \rightarrow x} g_n(t) &= \lim_{t \rightarrow x} \frac{f_n(t) - f_n(x)}{t - x} \\ &= f_n'(x), \quad n=1,2,3,\dots \end{aligned} \quad \dots(6)$$

$$\text{and } \lim_{t \rightarrow x} g(t) = \lim_{t \rightarrow x} \frac{f(t) - f(x)}{t - x} = f'(x) \quad \dots(7)$$

From(3), for any  $x, t$  in  $[a, b]$ ,  $m > N, n > N$  implies

$$|(f_n(x) - f_n(t)) - (f_m(x) - f_m(t))| \leq \frac{\epsilon}{2(b-a)} |x - t|$$

Therefore,

$$\begin{aligned} |g_n(t) - g_m(t)| &= \left| \frac{f_n(t) - f_n(x)}{t - x} - \frac{f_m(t) - f_m(x)}{t - x} \right| \\ &\leq \frac{\epsilon}{2(b-a)}, \text{ for all } t \text{ in } [a, b], t \neq x, m > N, n > N. \end{aligned}$$

Hence  $\{g_n\}$  satisfies Cauchy criterion for uniform convergence, for  $t \neq x$ .

Therefore  $\{f_n\}$  converges uniformly, for  $t \rightarrow x$ .

Also for  $t \rightarrow x$ ,

$$\begin{aligned} \lim_{n \rightarrow \infty} f_n(t) &= \lim_{n \rightarrow \infty} \frac{f_n(t) - f_n(x)}{t - x} \\ &= \frac{f(t) - f(x)}{t - x} \\ &= f'(t). \end{aligned}$$

Therefore,

$$\begin{aligned} \lim_{t \rightarrow x} f'(t) &= \lim_{t \rightarrow x} [\lim_{n \rightarrow \infty} f_n(t)] \\ &= \lim_{n \rightarrow \infty} [\lim_{t \rightarrow x} f_n(t)] && \text{[by theorem 6.3.1]} \\ &= \lim_{n \rightarrow \infty} f_n'(x) && \text{[from (6)]} \end{aligned}$$

Therefore,

$$\lim_{t \rightarrow x} f'(t) = \lim_{n \rightarrow \infty} f_n'(x) \quad \dots(8)$$

From (7) and (8), we get that

$$\begin{aligned} f'(x) &= \lim_{t \rightarrow x} f'(t) \\ &= \lim_{n \rightarrow \infty} f_n'(x). \end{aligned}$$

Hence the theorem.

### Check your progress

Is there exist a real continuous function on the line which is nowhere differentiable.

Answer:

Yes.

Define  $\phi(x) = |x|$  ( $-1 \leq x \leq 1$ )

Extend  $\phi$  to all real  $x$  by defining

$$\phi(x+2) = \phi(x).$$

Then  $\phi$  is continuous on  $\mathbb{R}^1$ .

Define  $f$  on  $\mathbb{R}^1$  as

$$f(x) = \sum_{n=0}^{\infty} \left(\frac{3}{4}\right)^n \phi(4^n x)$$

$f$  is the required function.

### 8.3 Let us sum up

In this lesson, we have seen that

- just uniform convergence alone is not sufficient for the limit of a sequence of differentiable functions to be differentiable and
- the conditions required for interchanging limit and differentiation.

### 8.4 Lesson End Activities

1. Show that the sequence  $\{f_n\}_{n=1}^{\infty}$  of functions where  $f_n(x) = nx / (1+n^2x^2)$ , converges to  $f$  where  $f(x) = 0 \forall x$  and that the equation  $\hat{f}(x) = \lim_{n \rightarrow \infty} f_n'(x)$  is true  $\forall x \neq 0$  but is false if  $x = 0$ .

### 8.5 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W. Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

**LESSON - 9****EQUICONTINUOUS FAMILY OF FUNCTIONS****9.0 Introduction****9.1 Aims and Objectives****9.2 Pointwise bounded functions****9.3 Uniformly bounded functions****9.4 Equicontinuous family of functions****9.5 Let us sum up****9.6 Lesson End Activities****9.7 References****9.0 Introduction**

In this lesson, we are going to study about pointwise bounded functions, uniformly bounded functions, equicontinuous family of functions.

**9.1 Aims and Objectives**

After studying this lesson, you should be able to identify

- Pointwise bounded functions
- Uniformly bounded functions
- Equicontinuous family of functions and
- Which of these families of functions contains a convergent subsequence.

**9.2 Pointwise bounded functions****Definition**

Let  $\{f_n\}$  be a sequence of functions defined on a set  $E$ .

$\{f_n\}$  is said to be pointwise bounded on  $E$ , if the sequence  $\{f_n(x)\}$  is bounded for every  $x \in E$ .

In other words,

$\{f_n\}$  is said to be pointwise bounded on  $E$ , if there exists a finite valued function defined on  $E$  such that

$$|f_n(x)| < (x) \quad x \in E, n=1,2,3,\dots$$

**Theorem 9.1**

If  $\{f_n\}$  is a pointwise bounded sequence of complex functions on a countable set  $E$ , then  $\{f_n\}$  has a subsequence  $\{f_{n_k}\}$  such that  $\{f_{n_k}(x)\}$  converges for every  $x \in E$ .

**Proof**

Let  $\{f_n\}$  be a pointwise bounded sequence of complex functions on a countable set  $E$ . Since  $E$  is countable,

The points of  $E$  can be arranged in a sequence as

$$\{x_i\}, i=1,2,3,\dots$$

Since  $\{f_n\}$  is pointwise bounded on  $E$ ,

$\{f_n(x_1)\}$  is a bounded sequence of complex numbers.

Therefore,

$\{f_n(x_1)\}$  contains a convergent subsequence.

Denote it by  $\{f_{1,k}\}$  i.e.  $\{f_{1,k}(x_1)\}$  converges as  $k \rightarrow \infty$ .

Now consider  $\{f_{1,k}(x_2)\}$ .

Since  $\{f_{1,k}\}$  is a subsequence of  $\{f_n\}$ ,

$\{f_{1,k}\}$  is also pointwise bounded on  $E$ .

Therefore,

$\{f_{1,k}(x_2)\}$  contains a convergent subsequence, say  $\{f_{2,k}(x_2)\}$ .

Proceeding in this way, we get the sequences represented by the following array:

$S_1: f_{1,1} \quad f_{1,2} \quad f_{1,3} \quad \dots$   
 $S_2: f_{2,1} \quad f_{2,2} \quad f_{2,3} \quad \dots$   
 $S_3: f_{3,1} \quad f_{3,2} \quad f_{3,3} \quad \dots$   
 $\dots$   
 $\dots$   
 $\dots$

which have the following properties:

- (1)  $S_n$  is a subsequence of  $S_{n-1}$ , for  $n=2,3,4,\dots$
- (2)  $\{f_{n,k}(x_n)\}$  converges as  $k \rightarrow \infty$
- (3) the order in which the functions appear is the same in each sequence. i.e if one function precedes another in  $S_1$ , then they are in same relation in every  $S_n$ , until one or the other is deleted.

Hence when going from one row in the above array to the next row below, functions may move left but never to the right.

Now consider the sequence

$S: f_{1,1} \quad f_{2,2} \quad f_{3,3} \quad f_{4,4} \quad \dots$

obtained by going down the diagonal of the array.

By (3), the sequence  $S$  is a subsequence of  $S_n$  (except possibly its first  $n-1$  terms), for  $n=1,2,3,\dots$

Hence (2) implies that

$\{f_{n,n}(x_i)\}$  converges for every  $x_i \in E$ .

Hence  $\{f_n\}$  has a subsequence that converges for every  $x_i \in E$ .

Hence the theorem.

### 9.3 Uniformly bounded functions

#### Definition:

Let  $\{f_n\}$  be a sequence of functions defined on a set  $E$ .  $\{f_n\}$  is said to be uniformly bounded on  $E$ , if there exists a Number  $M$  such that

$$|f_n(x)| < M \quad (x \in E, n=1,2,3,\dots)$$

#### Check your progress

1. In section 9.2, you have seen that pointwise bounded sequence of functions contains a convergent subsequence. Can you say that this will be true for uniformly bounded functions also?

2. Define  $f_n(x) = \frac{x^2}{x^2 + (1-nx)^2}$   $(0 \leq x \leq 1, n=1,2,3,\dots)$  Show that  $\{f_n\}$  is

uniformly bounded on  $[0, 1]$ .

3. Does every convergent sequence contain a uniformly convergent subsequence?

Answer

1. No.

Example

Let  $f_n(x) = \sin nx$   $(0 \leq x \leq 2\pi, n=1,2,3,\dots)$

Then  $|f_n(x)| = |\sin nx| \leq 1, (0 \leq x \leq 2\pi, n=1,2,3,\dots)$

Therefore,

$\{f_n\}$  is uniformly bounded on the compact set  $[0, 2\pi]$ .

Suppose there exists a sequence  $\{n_k\}$  such that

$\{\sin n_k x\}$  converges, for every  $x \in [0, 2\pi]$

Then  $\lim_{k \rightarrow \infty} (\sin n_k x - \sin n_{k+1} x) = 0 \quad 0 \leq x \leq 2\pi$ .

Therefore,

$$\lim_{k \rightarrow \infty} (\sin n_k x - \sin n_{k+1} x)^2 = 0 \quad 0 \leq x \leq 2\pi.$$

Therefore,

$$\lim_{k \rightarrow \infty} \int_0^{2\pi} (\sin n_k x - \sin n_{k+1} x)^2 dx \quad \dots(1)$$

But if we evaluate  $\int_0^{2\pi} (\sin n_k x - \sin n_{k+1} x)^2 dx$ , we get

$$\begin{aligned} \int_0^{2\pi} (\sin n_k x - \sin n_{k+1} x)^2 dx \\ = \int_0^{2\pi} (\sin^2 n_k x + \sin^2 n_{k+1} x - 2 \sin n_k x \sin n_{k+1} x) dx \\ = 2 \end{aligned}$$

Therefore,

$$\lim_{k \rightarrow \infty} \int_0^{2\pi} (\sin n_k x - \sin n_{k+1} x)^2 dx = 2, \text{ which contradicts (1).}$$

Therefore  $\{f_n\}$  does not have any convergent subsequence.

2.

Example:

Let  $f_n(x) = \frac{x^2}{x^2 + (1-nx)^2} \quad (0 \leq x \leq 1, n=1,2,3,\dots)$

Then  $|f_n(x)| \leq 1$ .

Hence  $\{f_n\}$  is uniformly bounded on  $[0, 1]$ .

3. Need not be so.

Let  $f_n(x) = \frac{x^2}{x^2 + (1-nx)^2} \quad (0 \leq x \leq 1, n=1,2,3,\dots)$

Then  $\lim_{n \rightarrow \infty} f_n(x) = 0 \quad 0 \leq x \leq 1$

i.e  $\{f_n\}$  converges pointwise to  $f(x)=0$  on  $[0, 1]$ .

But  $\{f_n\}$  does not contain any uniformly convergent subsequence, because

$$f_n(1/n) = 1 \text{ for all } n=1,2,3,\dots$$

**9.4 Equicontinuous family of functions**

**Definition**

A family  $F$  of complex functions  $f$  defined on a set  $E$  in a metric space  $X$  is said to be equicontinuous on  $E$  if, for every  $\epsilon > 0$ , there exists a  $\delta > 0$  such that

$$|f(x) - f(y)| < \epsilon, \text{ whenever } d(x,y) < \delta, x \in E, y \in E.$$

Here  $d$  denotes the metric of  $X$ .

**Check your progress**

1. Do you remember uniformly continuous function? Every member of an equicontinuous family of functions is uniformly continuous. True or False?

2. Let  $f_n(x) = \frac{x^2}{x^2 + (1-nx)^2}$  ( $0 \leq x \leq 1, n=1,2,3,\dots$ )

Is  $\{f_n\}$  is a equicontinuous family of functions?

Answer

1. True.
2. No.

**Theorem9.2**

If  $K$  is a compact metric space, if  $f_n \in C(K)$  for  $n=1,2,3,\dots$  and if  $\{f_n\}$  converges uniformly on  $K$ , then  $\{f_n\}$  is equicontinuous on  $K$ .

(Remark:

$C(K)$  denotes the set of all continuous functions on  $K$ .)

**Proof:**

Let  $\epsilon > 0$  be given.

Since  $\{f_n\}$  converges uniformly on  $K$ ,

$\{f_n\}$  satisfies the cauchy's condition for uniform convergence.

Hence for the given  $\epsilon > 0$ ,

there exists an integer  $N$  such that  $m \geq N, n \geq N, x \in K$  implies

$$|f_n(x) - f_m(x)| < \epsilon/3.$$

Hence,

$$|f_n(x) - f_N(x)| < \epsilon/3, \text{ whenever } n \geq N \text{ and } x \in K \quad \dots(1)$$

Hence,

$$\sup\{|f_n(x) - f_N(x)|, x \in K\} < \epsilon/3, \text{ whenever } n \geq N.$$

Hence

$$\|f_n - f_N\| < \epsilon/3, (n \geq N) \quad \dots(2)$$

Since continuous functions on a compact metric space are uniformly continuous and  $K$  is compact,

each  $f_i$  is a uniformly continuous function on  $K$ .

In particular  $f_1, f_2, f_3, \dots, f_N$  are uniformly continuous on  $K$ .

Therefore for the same  $\epsilon > 0$ ,

there exists a  $\delta > 0$  such that

$$|f_i(x) - f_i(y)| < \epsilon/3 \text{ whenever } d(x,y) < \delta \text{ and } i=1,2,\dots,N. \quad \dots(3)$$

From (1) and (3), we get that,

if  $n \geq N$  and  $d(x, y) < \delta$ ,

$$\begin{aligned} |f_n(x) - f_n(y)| &= |f_n(x) - f_N(x) + f_N(x) - f_N(y) + f_N(y) - f_n(y)| \\ &= |f_n(x) - f_N(x)| + |f_N(x) - f_N(y)| + |f_N(y) - f_n(y)| \\ &< \epsilon/3 + \epsilon/3 + \epsilon/3 = \epsilon \end{aligned} \quad \dots(4)$$

From (2) and (4),

$$|f_n(x) - f_n(y)| < \epsilon \text{ whenever } d(x, y) < \delta, x \in K, y \in K, n=1,2,3,\dots$$

Hence  $\{f_n\}$  is equicontinuous on  $K$ .

Hence the theorem.

**Theorem9.3**

If  $K$  is compact, if  $f_n \in C(K)$  for  $n=1,2,3,\dots$  and if  $\{f_n\}$  is pointwise bounded and equicontinuous on  $K$ , then

(a)  $\{f_n\}$  is uniformly bounded on  $K$

(b)  $\{f_n\}$  contains a uniformly convergent subsequence.

**Proof:**

Let  $K$  be compact and  $f_n \in C(K)$ ,  $n=1,2,3,\dots$

Let  $\{f_n\}$  is pointwise bounded and equicontinuous on  $K$ .

(a) To prove that  $\{f_n\}$  is uniformly bounded on  $K$ .

Since  $\{f_n\}$  is equicontinuous on  $K$ ,

for given  $\epsilon > 0$ , there exists a  $\delta > 0$  such that

$$|f_n(x) - f_n(y)| < \epsilon, \text{ whenever } d(x, y) < \delta, n=1,2,3,\dots \quad \dots(1)$$

Since  $K$  is compact, for this  $\delta > 0$ , there exists a finite subset  $\{p_1, p_2, \dots, p_r\}$  of  $K$  which is dense in  $K$ .

i.e for every  $x \in K$ , there corresponds atleast one  $p_i$  with

$$d(x, p_i) < \delta \quad \dots(2)$$

Since  $\{f_n\}$  is pointwise bounded on  $K$ ,

$\{f_n(p_i)\}$  is bounded for all  $p_i$ .

i.e there exist  $M_i$  such that

$$|f_n(p_i)| < M_i \quad i=1,2,\dots,r.$$

Let  $M = \text{Max}(M_1, M_2, \dots, M_r)$

Let  $x$  be any element of  $K$ .

Then by (2), there exists one  $p_i$  such that

$$d(x, p_i) < \delta$$

Then by (1),

$$|f_n(x) - f_n(p_i)| < \epsilon, \quad \text{for all } n.$$

Hence

$$\begin{aligned} |f_n(x)| &= |f_n(x) - f_n(p_i) + f_n(p_i)| \\ &= |f_n(x) - f_n(p_i)| + |f_n(p_i)| \\ &< \epsilon + M_i \quad \text{for all } n. \end{aligned}$$

Therefore  $\{f_n\}$  is uniformly bounded on  $K$ .

(b) To prove that  $\{f_n\}$  contains a uniformly convergent subsequence.

Since  $K$  is compact,  $K$  contains a countable dense subset, say  $E$ .

Then by the theorem 9.1 which states that,

“If  $\{f_n\}$  is a pointwise bounded sequence of complex functions on a countable set  $E$ , then  $\{f_n\}$  has a subsequence  $\{f_{n_k}\}$  such that  $\{f_{n_k}(x)\}$  converges for every  $x \in E$ .”,

$\{f_n\}$  has a subsequence  $\{f_{n_k}\}$  such that  $\{f_{n_k}(x)\}$  converges for every  $x \in E$ .

Let  $f_{n_k} = g_i$ , to simplify the notation.

Then  $\{g_i(x)\}$  converges uniformly for every  $x \in K$ .

To prove that  $\{g_i\}$  converges uniformly on  $K$ .

By (2), for every  $x \in K$ ,

there corresponds atleast one  $p_i$  with

$$d(x, p_i) < \delta.$$

Let  $V(x, \delta) = \{y \in K, d(x, y) < \delta\}$

Since  $E$  is dense in  $K$  and  $K$  is compact,

there are finitely many points  $x_1, x_2, \dots, x_m$  in  $E$  such that

$$K \subset V(x_1, \delta) \cup V(x_2, \delta) \cup \dots \cup V(x_m, \delta)$$

Since  $\{g_i(x)\}$  converges uniformly for every  $x \in E$ ,

$\{g_i(x_s)\}$  converges for  $s=1,2,\dots,m$ .

Hence for each  $s=1,2,\dots,m$ ,

$\{g_i(x_s)\}$  is a Cauchy sequence.

Hence for each  $s=1,2,\dots,m$ , and  $\epsilon > 0$ ,

there exists  $N_s$  such that

$$|g_i(x_s) - g_j(x_s)| < \epsilon, \text{ whenever } i \in N_s, j \in N_s.$$

Let  $N = \max(N_1, N_2, \dots, N_m)$

$$\text{Then } |g_i(x_s) - g_j(x_s)| < \epsilon, \text{ whenever } i \in N, j \in N \text{ and } s=1,2,\dots,m \quad \dots(3)$$

Let  $x$  be any point in  $K$ .

Then  $x \in V(x_s, \delta)$  for some  $s$ .

Hence  $d(x, x_s) < \delta$ .

Therefore by (2),

$$|g_i(x) - g_j(x_s)| < \epsilon, \text{ for every } i \in N, j \in N \quad \dots(4)$$

From (3) and (4), for  $i \in N, j \in N, x \in K$

$$\begin{aligned} |g_i(x) - g_j(x)| &= |g_i(x) - g_i(x_s) + g_i(x_s) - g_j(x_s) + g_j(x_s) - g_j(x)| \\ &\leq |g_i(x) - g_i(x_s)| + |g_i(x_s) - g_j(x_s)| + |g_j(x_s) - g_j(x)| \\ &< \epsilon + \epsilon + \epsilon = 3\epsilon. \end{aligned}$$

Therefore,

$\{g_i\}$  is uniformly convergent.

i.e  $\{f_{n_k}\}$  is uniformly convergent on  $K$ .

Hence  $\{f_n\}$  contains a uniformly convergent subsequence.

Hence the theorem.

### 9.5 Let us sum up

In this lesson we have studied about

- Pointwise bounded functions
- Uniformly bounded functions
- equicontinuous family of functions
- pointwise bounded sequence of functions on a countable set contains a convergent subsequence.
- Sequence of continuous functions on a compact metric space is equicontinuous.
- A sequence of pointwise bounded and equicontinuous functions on a compact space is uniformly bounded and contains a uniformly convergent subsequence.

### 9.6 Lesson End Activities

1. Prove that every uniformly convergent sequence of bounded functions is uniformly bounded.
2. Set  $f_n(x) = x^2 / x^2 + (1-nx)^2$ ;  $0 \leq x \leq 1$ ,  $n=1,2,3,\dots$  verify if the sequence of functions  $\{f_n\}_{n=1}^{\infty}$  is equi continuous.

### 9.7 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## Lesson 10 The Stone- Weierstrass Theorem

### 10.0 Introduction

#### 10.1 Aims and Objectives

#### 10.2 The Weierstrass Theorem

#### 10.3 Algebra of functions

#### 10.4 The Stone – Weierstrass theorem

#### 10.5 Let us sum up.

#### 10.6 Lesson End Activities

#### 10.7 References

### 10.0 Introduction

In this lesson, we are going to study about the original version of the Weierstrass theorem and the Stone's generalization of the Weierstrass Theorem.

#### 10.1 Aims and Objectives

After studying this lesson, you would know

- Weierstrass Theorem
- Algebra of functions
- Uniform closure of an algebra
- Separation of points and
- Stone-Weierstrass Theorem.

### 10.2 The Weierstrass Theorem

#### Theorem 10.1

If  $f$  is a continuous complex function on  $[a, b]$ , there exists a sequence of polynomials  $P_n$  such that

$$\lim_{n \rightarrow \infty} P_n(x) = f(x) \text{ uniformly on } [a, b].$$

If  $f$  is real, then  $P_n$  may be taken real.

#### Proof

Without loss of generality, assume that  $[a, b] = [0, 1]$ .

Also assume that  $f(0) = f(1) = 0$ .

Suppose if  $f(0) \neq 0$ ,  $f(1) \neq 0$ ,

$$\text{Consider } g(x) = f(x) - f(0) - x[f(1) - f(0)] \quad 0 \leq x \leq 1$$

Then  $g(0) = g(1) = 0$ .

Hence if  $g$  can be obtained as the limit of polynomials, then the same is true for  $f$  also.

Hence we will get the theorem.

Hence it is sufficient to prove the theorem with  $f(0) = f(1) = 0$ .

Let  $f$  be a continuous complex function on  $[0, 1]$  with  $f(0) = f(1) = 0$ .

Since a continuous function on a compact space is uniformly continuous,

$f$  is uniformly continuous on  $[0, 1]$ .

Define  $f(x)$  to be zero for  $x$  outside  $[0, 1]$ .

Then  $f$  is uniformly continuous on the whole real line.

$$\text{Define } Q_n(x) = c_n(1-x^2)^n, \quad n=1,2,3,\dots \quad \dots(1)$$

where  $c_n$  is chosen such that

$$\int_{-1}^1 Q_n(x) dx = 1, \quad n=1,2,3,\dots \quad \dots(2)$$

$$\text{i.e. } \int_{-1}^1 c_n(1-x^2)^n dx = 1, \quad n=1,2,3,\dots \quad \dots(3)$$

consider  $\int_{-1}^1 (1-x^2)^n dx = 2 \int_0^1 (1-x^2)^n dx$

$$= 2 \int_0^{1/\sqrt{n}} (1-x^2)^n dx$$

$$= 2 \int_0^{1/\sqrt{n}} (1-nx^2) dx$$

[since  $(1-x^2)^n = 1 - nx^2 + n(n-1)x^4/2 - \dots - nx^{2n}$ ]

$$= 2 \left[ x - n \frac{x^3}{3} \right]_0^{1/\sqrt{n}}$$

$$= \frac{4}{3\sqrt{n}}$$

$$> \frac{1}{\sqrt{n}} \tag{4}$$

From (3) and (4),

$$1 = \int_{-1}^1 c_n (1-x^2)^n dx > \frac{c_n}{\sqrt{n}}$$

Therefore  $c_n < \sqrt{n}$  ....(5)

Therefore for any  $\epsilon > 0$ ,

$$Q_n(x) = \frac{c_n (1-x^2)^n}{\sqrt{n}}$$

$$= \frac{c_n (1-x^2)^n}{\sqrt{n}} \quad \text{for } |x| \leq 1. \tag{6}$$

Therefore,

$$Q_n > 0 \text{ uniformly in } |x| \leq 1$$

Define  $P_n(x) = \int_{-1}^1 f(x+t) Q_n(t) dt$

Since f is zero outside [0, 1],

$$P_n(x) = \int_{-x}^{1-x} f(x+t) Q_n(t) dt$$

Changing the variable t as t-x, we get

$$P_n(x) = \int_0^1 f(t) Q_n(t-x) dt, \tag{7}$$

which is a polynomial in x.

Thus  $\{P_n\}$  is a sequence of polynomials, which are real if, f is real.

Since f is uniformly continuous,

for given  $\epsilon > 0$ , there exists a  $\delta > 0$  such that

$$|y-x| < \delta \text{ implies } |f(y) - f(x)| < \epsilon/2. \tag{8}$$

Let  $M = \sup |f(x)|$

For  $0 \leq x \leq 1$ ,

$$|P_n(x) - f(x)| = \left| \int_{-1}^1 f(x+t) Q_n(t) dt - f(x) \right|$$

$$= \left| \int_{-1}^1 f(x+t) Q_n(t) dt - \int_{-1}^1 f(x) Q_n(t) dt \right|$$

$$= \left| \int_{-1}^1 [f(x+t) - f(x)] Q_n(t) dt \right|$$

$$\leq \int_{-1}^1 |f(x+t) - f(x)| Q_n(t) dt \tag{since } Q_n(t) \geq 0$$

$$= \int_{-1}^{-\delta} |f(x+t) - f(x)| Q_n(t) dt + \int_{-\delta}^{\delta} |f(x+t) - f(x)| Q_n(t) dt + \int_{\delta}^1 |f(x+t) - f(x)| Q_n(t) dt$$

Now  $\frac{|f(x+t) - f(x)|}{2M} \leq \frac{|f(x+t)| + |f(x)|}{2M}$

Hence ,

$$\begin{aligned} |P_n(x) - f(x)| &\leq 2M \int_{-1}^{-\delta} Q_n(t) dt + \frac{\epsilon}{2} \int_{-\delta}^{\delta} Q_n(t) dt + 2M \int_{\delta}^1 Q_n(t) dt \\ &\leq 2M \sqrt{n} (1 - \delta^2)^n \int_{-1}^{-\delta} dt + \frac{\epsilon}{2} + 2M \sqrt{n} (1 - \delta^2)^n \int_{\delta}^1 dt \\ &\leq 2M \sqrt{n} (1 - \delta^2)^n (-\delta + 1) + \frac{\epsilon}{2} + 2M \sqrt{n} (1 - \delta^2)^n (1 - \delta) \\ &\leq 4M \sqrt{n} (1 - \delta^2)^n + \frac{\epsilon}{2} \\ &< \epsilon \quad \text{for sufficiently large } n. \\ &\quad \text{[since } \lim_{n \rightarrow \infty} (1 - \delta^2)^n = 0] \end{aligned}$$

Hence,

$$\lim_{n \rightarrow \infty} P_n(x) = f(x)$$

Hence the theorem.

**Theorem 10.2**

For every interval  $[-a, a]$  there is a sequence of real polynomials  $P_n$  such that  $\lim_{n \rightarrow \infty} P_n(x) = |x|$  uniformly on  $[-a, a]$ .

**Proof**

Since  $|x|$  is a continuous function on  $[-a, a]$ , by the above theorem 10.1, there exist a sequence  $\{P_n^*\}$  of real polynomials which converges to  $|x|$  uniformly on  $[-a, a]$ .

i.e  $\lim_{n \rightarrow \infty} P_n^*(x) = |x|$  uniformly on  $[-a, a]$ . ... (1)

Therefore ,

$$\lim_{n \rightarrow \infty} P_n^*(0) = 0 \quad \text{... (2)}$$

Let  $P_n(x) = P_n^*(x) - P_n^*(0)$

Then from (1) and (2),

$$P_n(0) = P_n^*(0) - P_n^*(0) = 0.$$

and  $\lim_{n \rightarrow \infty} P_n(x) = \lim_{n \rightarrow \infty} [P_n^*(x) - P_n^*(0)]$   
 $= |x| - 0$   
 $= |x|$  uniformly on  $[-a, a]$ .

Hence the theorem.

**10.3 Algebra of functions**

**Definition**

A family  $\mathcal{A}$  of complex functions defined on a set  $E$  is said to be an algebra if

- (i)  $f+g \in \mathcal{A}$
- (ii)  $fg \in \mathcal{A}$  and
- (iii)  $cf \in \mathcal{A}$

for all  $f \in \mathcal{A}, g \in \mathcal{A}$  and for all complex constants  $c$ .

i.e if  $\mathcal{A}$  is closed under addition, multiplication and scalar multiplication.

**Remark**

If we consider algebras of real functions,  
then (iii) is required to hold for all real  $c$ .

**Definition**

If  $A$  has the property that  $f \in A$  whenever

- (i)  $f_n \in A$  ( $n=1,2,3,\dots$ ) and
  - (ii)  $f_n \rightarrow f$  uniformly on  $E$ ,
- then  $A$  is said to be uniformly closed.

**Definition**

Let  $B$  be the set of all functions which are limits of uniformly convergent sequences of members of  $A$ . Then  $B$  is called uniform closure of  $A$ .

**Theorem 10.3**

Let  $B$  be the uniform closure of an algebra  $A$  of bounded functions.

Then  $B$  is a uniformly closed algebra.

**Proof**

Let  $f \in B$  and  $g \in B$ .

Then there exists uniformly convergent sequences of functions  $\{f_n\}$  and  $\{g_n\}$  such that

$f_n \rightarrow f$  uniformly and  $g_n \rightarrow g$  uniformly.

Hence  $f_n + g_n \rightarrow f + g$  uniformly,

$f_n g_n \rightarrow fg$  uniformly

and  $cf_n \rightarrow cf$  uniformly, where  $c$  is a constant.

Hence

$f + g \in A$ ,

$fg \in A$  and

$cf \in A$ .

Hence  $B$  is uniformly closed.

**Definition**

Let  $A$  be a family of functions on a set  $E$ .

Then  $A$  is said to separate points on  $E$  if, to every pair of distinct points  $x_1, x_2 \in E$ , there corresponds a function  $f \in A$  such that

$$f(x_1) \neq f(x_2).$$

**Definition**

Let  $A$  be a family of functions on a set  $E$ .  $A$  is said to vanish at no point of  $E$  if, to each  $x \in E$ , there corresponds a function  $g \in A$  such that  $g(x) \neq 0$ .

**Check your progress**

1. Can you give an example of an algebra of functions which separates points and vanishes at no point.
2. Can you give an example of algebra of functions which does not separate points.

**Answer**

1. The algebra of polynomials in one variable has these properties on  $\mathbb{R}^1$ .

2. The set of all even polynomials on  $[-1, 1]$ .

**Theorem 10.4**

Suppose  $A$  is an algebra of functions on a set  $E$ ,  $A$  separates points on  $E$ , and  $A$  vanishes at no point of  $E$ .

Suppose  $x_1, x_2$  are distinct points of  $E$  and  $c_1, c_2$  are constants (real if  $A$  is a real algebra). Then  $A$  contains a function  $f$  such that

$$f(x_1) = c_1, f(x_2) = c_2$$

**Proof**

Let  $x_1, x_2$  be distinct points of  $E$ .

Since  $A$  separates points on  $E$ ,

$A$  contains functions  $g, h, k$  such that

$$\begin{aligned} g(x_1) &= g(x_2), \\ h(x_1) &= 0, \\ k(x_2) &= 0. \end{aligned} \quad \dots(1)$$

Let  $u = gk - g(x_1)k$  and  $v = gh - g(x_2)h$ .

Since  $A$  is an algebra,

$$g, h, k \in A \text{ implies } u \in A, v \in A.$$

Also  $u(x_1) = g(x_1)k(x_1) - g(x_1)k(x_1) = 0$ ,

$$v(x_2) = g(x_2)h(x_2) - g(x_2)h(x_2) = 0,$$

$$u(x_2) = g(x_2)k(x_2) - g(x_1)k(x_2)$$

$$= [g(x_2) - g(x_1)]k(x_2)$$

$$= 0$$

[from (1)]

$$v(x_1) = g(x_1)h(x_1) - g(x_2)h(x_1)$$

$$= [g(x_2) - g(x_1)]h(x_1)$$

$$= 0$$

[from (1)]

Define  $f$  on  $E$  as

$$f = \frac{c_1 v}{v(x_1)} + \frac{c_2 u}{u(x_2)}.$$

Since  $u, v \in A$  and  $A$  is an algebra,

$$f \in A.$$

And 
$$f(x_1) = \frac{c_1 v(x_1)}{v(x_1)} + \frac{c_2 u(x_1)}{u(x_2)}$$

$$= c_1 + 0$$

$$= c_1,$$

$$f(x_2) = \frac{c_1 v(x_2)}{v(x_1)} + \frac{c_2 u(x_2)}{u(x_2)}$$

$$= 0 + c_2$$

$$= c_2.$$

Hence  $f$  has the desired properties.

Hence the theorem.

### 10.4 The Stone – Weierstrass theorem

#### Theorem 10.5 Stone's generalization of the theorem.

Let  $A$  be an algebra of real continuous functions on a compact set  $K$ . If  $A$  separates points on  $K$  and if  $A$  vanishes at no point of  $K$ , then the uniform closure  $B$  of  $A$  consists of all real continuous functions on  $K$ .

#### Proof

We shall divide the proof into four steps.

Step1: If  $f \in B$ , then  $|f| \in B$ .

Proof:

$$\text{Let } a = \sup\{|f(x)|, x \in K\} \quad \dots(1)$$

Let  $\epsilon > 0$  be given.

Then by the theorem 10.2, which states that,

“For every interval  $[-a, a]$  there is a sequence of real polynomials  $P_n$  such that  $\lim_n P_n(x) = |x|$  uniformly on  $[-a, a]$ .”

there exists a sequence of real polynomials  $\{P_n\}$  such that

$$P_n(0) = 0 \text{ and } \lim_n P_n(y) = |y| \text{ uniformly on } [-a, a].$$

Therefore,

there exists real numbers  $c_1, c_2, \dots, c_n$  such that

$$\left| \sum_{i=1}^n c_i y^i - |y| \right| < \epsilon \quad -a \leq y \leq a \quad \dots(2)$$

Since  $B$  is an algebra,

$$\text{the function } g = \sum_{i=1}^n c_i f^i \text{ is a member of } B.$$

Therefore,

$$\begin{aligned} |g(x) - |f(x)|| &= \left| \sum_{i=1}^n c_i f^i(x) - |f(x)| \right| \\ &= \left| \sum_{i=1}^n c_i y^i - |y| \right|, \quad \text{where } y = f(x). \end{aligned} \quad \dots(3)$$

Since  $a = \sup\{|f(x)|, x \in K\}$

$$-a \leq f(x) \leq a, \quad \text{for all } x \in K.$$

i.e.  $-a \leq y \leq a$ .

Hence by(2),

$$\left| \sum_{i=1}^n c_i y^i - |y| \right| < \epsilon \quad \dots(4)$$

From (3) and (4),

$$|g(x) - |f(x)|| < \epsilon \quad \text{for all } x \in K.$$

This implies that

$$|f| \in B, \quad \text{since } B \text{ is uniformly closed.}$$

Hence the claim that,

$$\text{if } f \in B, \text{ then } |f| \in B.$$

Step 2: If  $f \in B$  and  $g \in B$ , then  $\max(f,g) \in B$  and  $\min(f,g) \in B$ .

Proof:

By definition,

$$\max(f, g)(x) = \begin{cases} f(x) & \text{if } f(x) \geq g(x) \\ g(x) & \text{if } f(x) < g(x) \end{cases}$$

Therefore,

$$\max(f, g) = \frac{f + g}{2} + \frac{|f - g|}{2}$$

$$\text{Similarly } \min(f, g) = \frac{f + g}{2} - \frac{|f - g|}{2}$$

Since  $f \in B$ ,  $g \in B$ , and  $B$  is closed,

$$f + g \in B$$

$$\text{and } f - g \in B.$$

By the proof of step 1,

$$|f - g| \in B.$$

Hence,

$$\max(f, g) = \frac{f + g}{2} + \frac{|f - g|}{2} \in B$$

$$\text{and } \min(f, g) = \frac{f + g}{2} - \frac{|f - g|}{2} \in B.$$

Hence step 2 is proved.

By repeating this, we get that if  $f_1, f_2, \dots, f_n \in B$ ,

$$\text{then } \max(f_1, f_2, \dots, f_n) \in B$$

$$\text{and } \min(f_1, f_2, \dots, f_n) \in B$$

Step 3: Given a real valued function  $f$ , continuous on  $K$ , a point  $x \in K$  and  $\epsilon > 0$ , there exists a function  $g_x \in B$  such that

$$\begin{aligned} g_x(x) &= f(x) \\ \text{and } g_x(t) &> f(t) - \epsilon \quad \text{for all } t \in K. \end{aligned} \quad \dots(5)$$

Proof:

Let  $x$  be the given point in  $K$ .

Then for every  $y$  in  $K$  with  $y \neq x$ , by theorem 10.4,

taking  $c_1 = f(x)$  and  $c_2 = f(y)$ , we get that,

there exists a function  $h_y \in A$  satisfying

$$\begin{aligned} h_y(x) &= f(x) \text{ and} \\ h_y(y) &= f(y). \end{aligned} \quad \dots(6)$$

Since  $A \subset B$ ,  $h_y \in B$ .

Since  $h_y$  is continuous,

there exists an open set  $J_y$  containing  $y$  such that

$$h_y(t) > f(t) - \epsilon \quad \text{for all } t \in J_y \quad \dots(7)$$

Then  $K \subset \bigcup_{y \in K} J_y$ .

Since  $K$  is compact,

there is a finite set of points  $y_1, y_2, \dots, y_n$  such that

$$K \subset J_{y_1} \cup J_{y_2} \cup \dots \cup J_{y_n}.$$

Let  $g_x = \max(h_{y_1}, h_{y_2}, \dots, h_{y_n})$

By step 2,  $g_x \in \mathcal{B}$ .

Now from (6),

$$h_{y_i}(x) = f(x) \text{ for all } i=1,2,\dots,n.$$

Therefore,

$$g_x(x) = \max(h_{y_1}, h_{y_2}, \dots, h_{y_n})(x) = f(x).$$

Also from (7),

$$h_{y_i}(t) > f(t) - \epsilon, \text{ for all } i=1,2,\dots,n.$$

Therefore,

$$\begin{aligned} g_x(t) &= \max(h_{y_1}, h_{y_2}, \dots, h_{y_n})(t) \\ &> f(t) - \epsilon, \text{ for all } t \in K. \end{aligned}$$

Hence  $g_x$  has the desired properties.

Hence step 3 is proved.

Step 4:

Given a real valued function  $f$  continuous on  $K$  and  $\epsilon > 0$ , there exists a function  $h \in \mathcal{B}$  such that

$$|h(x) - f(x)| < \epsilon \text{ for all } x \in K.$$

i.e the uniform closure  $\mathcal{B}$  of  $\mathcal{A}$  consists of all real continuous functions on  $K$ .

Proof:

Consider the functions  $g_x$ , for each  $x \in K$  constructed in step 3.

Since each  $g_x$  is continuous on  $K$ ,

there exist open sets  $V_x$  containing  $x$  such that

$$g_x(t) < f(t) + \epsilon \text{ for all } t \in V_x \quad \dots(8)$$

Then  $K \subset \bigcup_{x \in K} V_x$ .

Since  $K$  is compact,

there exists a finite set of points  $x_1, x_2, \dots, x_m$  such that

$$K \subset V_{x_1} \cup V_{x_2} \cup \dots \cup V_{x_m}$$

Let  $h = \min(g_{x_1}, g_{x_2}, \dots, g_{x_m})$

Then  $h \in \mathcal{B}$ , by step 2.

By (5),  $g_{x_i}(t) > f(t) - \epsilon$ , for all  $t \in K, i=1,2,\dots,n$ .

Therefore,

$$h(t) > f(t) - \epsilon, \text{ for all } t \in K. \quad \dots(9)$$

By (8),  $g_{x_i}(t) < f(t) + \epsilon$ , for all  $t \in K, i=1,2,\dots,n$

Therefore,

$$h(t) < f(t) + \epsilon, \text{ for all } t \in K. \quad \dots(10)$$

From (9) and (10),

$$f(t) - \epsilon < h(t) < f(t) + \epsilon, \text{ for all } t \in K.$$

Hence,

$$|h(t) - f(t)| < \epsilon, \text{ for all } t \in K.$$

Hence for every real function  $f$  continuous on  $K$  and  $\epsilon > 0$ , there exists a function  $h \in \mathcal{B}$  such that

$$|h(t) - f(t)| < \epsilon, \text{ for all } t \in K.$$

Hence the uniform closure  $\bar{B}$  of  $A$  consists of all real continuous functions on  $K$ .  
Hence the theorem.

### 10.5 Let us sum up

In this lesson, we have seen

- The original version of Weierstrass Theorem which says that every continuous function (real or complex) on a finite interval is the limit of a sequence of polynomials.
- Definitions of algebra of functions and Uniform closure of an algebra
- Definition of Separation of points and Separation theorem and
- The Stone-Weierstrass Theorem and its proof.

### 10.6 Lesson End Activities

1. if  $f$  is continuous on  $[0, 1]$  and if  $\int_0^1 f(x) x^n dx = 0$  ( $n=0,1,2,\dots$ ). Prove that  $f(x) = 0$  on  $[0,1]$ .
2. let  $K$  be the unit circle in the complex plane and set  $A$  be the algebra of all function of the form.

$N$

$$F(e^{i\theta}) = \sum_{n \in \mathbb{N}} C_n e^{in\theta} \quad (\theta \text{ real})$$
 then prove that  $A$  separates points on  $K$  and that  $A$  vanishes at no point of  $K$  but there are continuous functions on  $K$  which are not in the uniform closure of  $A$ .

### 10.7 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Wiley and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

**UNIT III**  
**FUNCTIONS OF SEVERAL VARIABLES**  
**LESSON – 11**  
**LINEAR TRANSFORMATIONS**

**Contents**

- 11.0 Introduction**
- 11.1 Aims and objectives**
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**11.0 Introduction**

In this lesson, first we are going to be introduced to the definition of a vector space and some of its properties, which you might have studied in UG classes and then we are going to see about linear transformations, linear operators and the matrix representation of a linear transformation.

**11.1 Aims and objectives**

After studying this lesson you would know

- What is a linear transformation?
- What is a linear operator?
- Inverse of an operator.
- Some of the properties of Vector spaces, linear transformations and inverse operators and
- How to represent a linear transformation in the matrix form?

**11.2 Vector Spaces**

**Definition**

A nonempty set  $X \subset \mathbb{R}^n$  is said to be a vector space if,

- (i)  $x + y \in X$  and
- (ii)  $cx \in X$  for all  $x \in X, y \in X$  and for all scalars  $c$ .

**Definition**

If  $x_1, x_2, \dots, x_k \in \mathbb{R}^n$  and  $c_1, c_2, \dots, c_k$  are scalars, the vector  $c_1x_1 + c_2x_2 + \dots + c_kx_k$  is called a linear combination of  $x_1, x_2, \dots, x_k$ .

**Definition**

If  $S \subset \mathbb{R}^n$  and if E is the set of all linear combinations of elements of S, then we say that S spans E or that E is the span of S.

**Definition**

A set consisting of vectors  $x_1, x_2, \dots, x_k$  is said to be independent if the relation  $c_1x_1 + c_2x_2 + \dots + c_kx_k = 0$  implies that  $c_1 = c_2 = \dots = c_k = 0$ . Otherwise  $\{x_1, x_2, \dots, x_k\}$  is said to be linearly dependent.

**Definition**

If a vector space X contains an independent set of r vectors but contains no independent set of r+1 vectors, then we say that X has dimension r. We write this as  $\dim X = r$ .

**Definition**

An independent subset of a vector space X which spans X is called a basis of X.

Remarks:

1. The set consisting of 0 alone is a vector space and its dimension is 0.
2. If  $B = \{x_1, x_2, \dots, x_r\}$  is a basis of X, then B spans X and B is independent.

Hence every  $x \in X$  has a unique representation of the form

$$x = c_1x_1 + c_2x_2 + \dots + c_nx_n.$$

The numbers  $c_1, c_2, \dots, c_r$  are called the coordinates of X with respect to the basis B.

**Example:**

The set  $\{e_1, e_2, \dots, e_n\}$ , where  $e_j$  is the vector in  $\mathbb{R}^n$  whose jth coordinate is 1 and whose other coordinates are all 0.

i.e  $e_1 = (1, 0, 0, \dots, 0)$

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

.....

.....

and  $e_n = (0, 0, 0, \dots, 1)$

Now, if  $x = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$ ,

then ,

$$\begin{aligned} x &= x_1(1, 0, 0, \dots, 0) + x_2(0, 1, 0, \dots, 0) + \dots + x_n(0, 0, 0, \dots, 1) \\ &= x_1e_1 + x_2e_2 + \dots + x_n e_n. \end{aligned}$$

Hence  $\{e_1, e_2, \dots, e_n\}$  forms a basis of  $\mathbb{R}^n$ .

The set  $\{e_1, e_2, \dots, e_n\}$  is called the standard basis of  $\mathbb{R}^n$ .

**Some important results in vector spaces:**

1. Let r be a positive integer. If a vector space X is spanned by a set of r vectors, then  $\dim X = r$ .
2.  $\dim \mathbb{R}^n = n$ .

3. Suppose  $X$  is a vector space, and  $\dim X = n$ .
- A set  $E$  of  $n$  vectors in  $X$  spans  $X$  if and only if  $E$  is independent.
  - $X$  has a basis and every basis consists of  $n$  vectors.
  - If  $1 \leq r \leq n$  and  $\{y_1, y_2, \dots, y_r\}$  is an independent set in  $X$ , then  $X$  has a basis containing  $\{y_1, y_2, \dots, y_r\}$ .

### Check your progress

1. Every span of a set is a vector space. True or False?

2.  $\dim \mathbb{R}^3 = ?$

## 11.3 Linear transformations and linear operators

### Definition

A mapping  $A$  of a vector space  $X$  into a vector space  $Y$  is said to be a linear transformation if

- $A(x_1 + x_2) = A(x_1) + A(x_2)$  and
- $A(cx) = cA(x)$ , for all  $x, x_1, x_2 \in X$  and all scalars  $c$ .

\*If  $A$  is Linear, then  $A(x)$  is simply written as  $Ax$ .

\*If  $A$  is linear  $A0=0$ .

### Remark:

A linear transformation  $A$  of  $X$  into  $Y$  is completely determined by its action on any basis.

### Proof

If  $\{x_1, x_2, \dots, x_n\}$  is a basis of  $X$ ,

then every  $x \in X$  has a unique representation of the form

$$x = \sum_{i=1}^n c_i x_i.$$

Since  $A$  is linear,

$$\begin{aligned} Ax &= A\left(\sum_{i=1}^n c_i x_i\right) \\ &= \sum_{i=1}^n c_i Ax_i. \end{aligned}$$

Hence  $Ax$  can be computed from the vectors  $Ax_1, Ax_2, \dots, Ax_n$  and the coordinates  $c_1, c_2, \dots, c_n$ .  
i.e.  $R(A)$  is the span of the set  $Q = \{Ax_1, Ax_2, \dots, Ax_n\}$

### Definition

Linear transformations of  $X$  into  $X$  are called linear operators on  $X$ .

### Definition

If  $A$  is a linear operator on  $X$  such that

- $A$  is one-to-one and
- $A$  maps  $X$  onto  $X$ ,

then  $A$  is said to be invertible and

the operator  $A^{-1}$  on  $X$  is defined such that

$$A^{-1}(Ax) = x \text{ for all } x \in X.$$

### Check your progress

1. Prove that  $A(A^{-1}x) = x$  for all  $x \in X$ .

2. Prove that  $A^{-1}$  is linear

**Theorem 11.1**

A linear operator  $A$  on a finite dimensional vector space  $X$  is one-to-one if and only if the range of  $A$  is all  $X$ .

**Proof**

Let  $X$  be a vector space of dimension  $n$  and

$\{x_1, x_2, \dots, x_n\}$  be a basis of  $X$ .

Since  $A$  is linear,

$R(A)$  is the span of the set  $Q = \{Ax_1, Ax_2, \dots, Ax_n\}$ .

We know that

“A set  $E$  of  $n$  vectors in  $X$  spans  $X$  if and only if  $E$  is independent”.

Hence  $R(A) = X$  if and only if  $Q$  is independent.

Hence, it is enough to prove that  $Q$  is independent if and only if  $A$  is one-to-one.

To prove that  $Q$  is independent if and only if  $A$  is one-to-one.

Assume that  $A$  is one-to-one.

Suppose  $\sum_{i=1}^n c_i Ax_i = 0$ ,

Then,

$$A\left(\sum_{i=1}^n c_i x_i\right) = \sum_{i=1}^n c_i Ax_i = 0. \text{ [since } A \text{ is linear]}$$

Since  $A$  is one-to-one, this implies that

$$\sum_{i=1}^n c_i x_i = 0$$

Since  $\{x_1, x_2, \dots, x_n\}$  is linearly independent, this implies that

$$c_1 = c_2 = \dots = c_n = 0.$$

Hence  $Q = \{Ax_1, Ax_2, \dots, Ax_n\}$  is linearly independent.

Conversely, assume that  $Q$  is linearly independent.

Suppose if,

$$A\left(\sum_{i=1}^n c_i x_i\right) = 0.$$

Then  $\sum_{i=1}^n c_i Ax_i = 0$  [since  $A$  is linear]

This implies that  $c_1 = c_2 = \dots = c_n = 0$ . [since  $Q$  is independent]

This implies that  $x = \sum_{i=1}^n c_i x_i = 0$

i.e.  $Ax = 0$  only if  $x = 0$  ... (1)

Therefore,

if  $Ax = Ay$ ,

Then  $A(x-y) = Ax - Ay = 0$

Then by (1),  $x-y=0$ .

Hence  $x=y$ .

Hence  $A$  is one-to-one.

Hence the result

Hence  $A$  is one-to-one if and only if  $R(A) = X$ .

Hence the theorem.

### Definition

$L(X, Y)$  denote the set of all linear transformations of the vector space  $X$  into the vector space  $Y$ .

$L(X, X)$  can be simply written as  $L(X)$ .

i.e  $L(X)$  denotes the set of all operators on  $X$ .

### Definition

If  $A_1, A_2 \in L(X, Y)$  and if  $c_1, c_2$  are scalars,  
 $c_1A_1 + c_2A_2$  is defined by

$$(c_1A_1 + c_2A_2)(x) = c_1A_1x + c_2A_2x, \quad \text{for } x \in X.$$

Since  $A_1$  and  $A_2$  are linear,  $c_1A_1 + c_2A_2$  is also linear.

Hence  $c_1A_1 + c_2A_2 \in L(X, Y)$

### Definition

If  $X, Y, Z$  are vector spaces and

if  $A \in L(X, Y), B \in L(Y, Z)$ ,

Their product  $BA$  is defined to be the composition of  $A$  and  $B$ .

i.e  $(BA)(x) = B(Ax), \quad \text{for all } x \in X.$

Then  $BA \in L(X, Y)$

### Definition

For  $A \in L(\mathbb{R}^n, \mathbb{R}^m)$ , the norm of  $A$  is defined as

$$\|A\| = \sup \{ |A(x)|, x \in \mathbb{R}^n \text{ with } |x| = 1 \}$$

Remark:

$$|A(x)| \leq \|A\| |x|, \quad \text{for all } x \in \mathbb{R}^n$$

Proof

Let  $x \in \mathbb{R}^n$ .

Define  $y = \frac{x}{|x|}$

Then  $|y| = 1$ .

Therefore by definition of  $\|A\|$ ,

$$|A(y)| \leq \|A\|$$

Therefore ,

$$|A\left(\frac{x}{|x|}\right)| \leq \|A\|$$

Therefore ,

$$\frac{|A(x)|}{|x|} \leq \|A\|$$

Therefore ,

$$|A(x)| \leq \|A\| |x|, \quad \text{for all } x \in \mathbb{R}^n.$$

Hence the result.

Also, if  $x$  is such that  $|A(x)| = |x|$ , for all  $x \in \mathbb{R}^n$ ,  
 then  $\|A\| = 1$ .

### Check your progress

1. If  $A_1, A_2 \in L(X, Y)$  and if  $c_1, c_2$  are scalars, prove that  $c_1A_1 + c_2A_2$  is linear i.e  $c_1A_1 + c_2A_2 \in L(X, Y)$
2. If  $A \in L(X, Y), B \in L(Y, Z)$ , prove that  $BA \in L(X, Y)$
3. If  $X = Y = Z$ , is  $BA = AB$ ?

**Answers**

1. For  $x_1, x_2 \in X$  and scalars  $a, b$ , consider

$$\begin{aligned} (c_1A_1 + c_2A_2)(ax_1 + bx_2) &= c_1A_1(ax_1 + bx_2) + c_2A_2(ax_1 + bx_2) \\ &\quad \text{[by definition of } c_1A_1 + c_2A_2\text{]} \\ &= c_1(aA_1x_1 + bA_1x_2) + c_2(aA_2x_1 + bA_2x_2) \\ &= c_1aA_1x_1 + c_1bA_1x_2 + c_2aA_2x_1 + c_2bA_2x_2 \\ &= a(c_1A_1 + c_2A_2)(x_1) + b(c_1A_1 + c_2A_2)(x_2) \end{aligned}$$

Hence  $c_1A_1 + c_2A_2$  is linear.

2. Prove it by yourself.

3.  $BA$  need not be the same as  $AB$ .

**Theorem 11.2**

(a) If  $A \in L(\mathbb{R}^n, \mathbb{R}^m)$ , then  $\|A\| < \infty$  (i.e  $\|A\|$  is finite) and  $A$  is a uniformly continuous mapping of  $\mathbb{R}^n$  into  $\mathbb{R}^m$ .

(b) If  $A, B \in L(\mathbb{R}^n, \mathbb{R}^m)$  and  $c$  is a scalar, then  $\|A+B\| \leq \|A\| + \|B\|$ ,  $\|cA\| = |c| \|A\|$ .

If we define distance between  $A$  and  $B$  as  $\|A - B\|$ , then  $L(\mathbb{R}^n, \mathbb{R}^m)$  is a metric space.

(c) If  $A \in L(\mathbb{R}^n, \mathbb{R}^m)$  and  $B \in L(\mathbb{R}^n, \mathbb{R}^m)$ , then  $\|BA\| \leq \|B\| \|A\|$ .

**Proof**

(a) Let  $A \in L(\mathbb{R}^n, \mathbb{R}^m)$ .

Let  $\{e_1, e_2, \dots, e_n\}$  be the basis in  $\mathbb{R}^n$ .

Suppose that  $x = \sum_{i=1}^n c_i e_i$  with  $\|x\| = 1$ .

Then  $|c_i| \leq 1$  for all  $i=1, 2, \dots, n$ .

Therefore,

$$\begin{aligned} \|A(x)\| &= \left\| A\left(\sum_{i=1}^n c_i e_i\right) \right\| \\ &= \left\| \sum_{i=1}^n c_i A e_i \right\| \\ &\leq \sum_{i=1}^n |c_i| \|A e_i\| \\ &\leq \sum_{i=1}^n \|A e_i\| \end{aligned}$$

Therefore,

$$\sup \{ \|A(x)\|, x \in \mathbb{R}^n \text{ with } \|x\| = 1 \} = \sum_{i=1}^n \|A e_i\| < \infty$$

i.e  $\|A\| < \infty$ .

To prove that  $A$  is uniformly continuous

From the inequality  $|A(x)| \leq \|A\| |x|$ , for all  $x \in \mathbb{R}^n$ ,

$$|Ax - Ay| \leq \|A\| |x - y|, \text{ for all } x, y \in \mathbb{R}^n.$$

Therefore for any  $\epsilon > 0$ ,

if we choose  $\delta = \frac{\epsilon}{\|A\|}$  (independent of  $x$  and  $y$ ), then

$$|x - y| < \delta \text{ implies that } |Ax - Ay| \leq \|A\| |x - y| < \epsilon.$$

Hence for any  $\epsilon > 0$ , there exists a  $\delta > 0$  such that

$$|Ax - Ay| < \epsilon \text{ whenever } |x - y| < \delta, x, y \in \mathbb{R}^n.$$

Hence  $A$  is uniformly continuous.

(b) To prove that, if  $A, B \in L(\mathbb{R}^n, \mathbb{R}^m)$  and  $c$  is a scalar,

then  $\|A+B\| = \|A\| + \|B\|$  and  $\|cA\| = |c| \|A\|$ .

$$\begin{aligned} \text{Consider } |(A+B)(x)| &= |Ax+Bx| \\ &= |Ax| + |Bx| \\ &\leq \|A\| |x| + \|B\| |x| \end{aligned}$$

Therefore,

$$\text{if } |x| = 1,$$

$$\text{Then } |(A+B)(x)| \leq \|A\| + \|B\|$$

Therefore,

$$\sup\{|(A+B)(x)|, x \in \mathbb{R}^n, |x| = 1\} \leq \|A\| + \|B\|$$

Therefore,

$$\|A+B\| \leq \|A\| + \|B\|$$

$$\begin{aligned} \text{Consider } |(cA)(x)| &= |c(Ax)| \\ &= |c| |Ax|. \end{aligned}$$

Therefore,

$$\sup\{|(cA)(x)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} = |c| \sup\{|Ax|, x \in \mathbb{R}^n \text{ and } |x| = 1\}$$

Therefore,

$$\|cA\| = |c| \|A\|.$$

To prove that  $L(\mathbb{R}^n, \mathbb{R}^m)$  is a metric space.

Define the metric on  $L(\mathbb{R}^n, \mathbb{R}^m)$  as follows:

If  $A, B \in L(\mathbb{R}^n, \mathbb{R}^m)$ ,

define the distance between  $A$  and  $B$  as  $\|A - B\|$ .

Then

$$1. \quad \|A-B\| = \sup\{|(A-B)(x)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} \geq 0$$

$$\begin{aligned} 2. \quad \|A-B\| = 0 &\text{ if and only if } \sup\{|(A-B)(x)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} = 0 \\ &\text{if and only if } |(A-B)(x)| = 0, \text{ for all } x \in \mathbb{R}^n \text{ with } |x| = 1. \\ &\text{if and only if } |Ax - Bx| = 0, \text{ for all } x \in \mathbb{R}^n \text{ with } |x| = 1 \\ &\text{if and only if } Ax = Bx, \text{ for all } x \in \mathbb{R}^n \text{ with } |x| = 1 \quad \dots(1) \end{aligned}$$

Suppose if  $y \in \mathbb{R}^n$ , with  $y \neq 0$ ,

$$\text{then } x = \frac{y}{|y|} \text{ is such that } |x| = 1$$

$$\text{Therefore, } Ax = Bx$$

$$\text{implies } A\left(\frac{y}{|y|}\right) = B\left(\frac{y}{|y|}\right)$$

implies 
$$\frac{1}{|y|} Ay = \frac{1}{|y|} By$$

implies  $Ay = By$  [Since  $|y| \neq 0$ , we can divide by  $|y|$ ]

Using this in (1), we get that

$$\|A-B\| = 0 \text{ if and only if } Ay=By, \text{ for all } y \in \mathbb{R}^n.$$

Hence  $\|A-B\| = 0$  if and only if  $A=B$ .

$$\begin{aligned} 3. \quad \|A-B\| &= \sup\{|(A-B)(x)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} \\ &= \sup\{|(B-A)(x)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} \\ &= \|B-A\| \end{aligned}$$

$$\begin{aligned} 4. \quad \|A-B\| &= \sup\{|(A-B)(x)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} \\ &= \sup\{|(Ax-Bx)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} \\ &= \sup\{|(Ax-Cx + Cx-Bx)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} \end{aligned}$$

$$\begin{aligned} &\sup\{|(Ax-Cx)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} \\ &\quad + \sup\{|(Cx-Bx)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} \\ &= \|A-C\| + \|C-B\| \end{aligned}$$

Hence the triangle inequality is satisfied.

Hence the metric we defined on  $L(\mathbb{R}^n, \mathbb{R}^m)$  satisfies all the required properties.

Hence  $L(\mathbb{R}^n, \mathbb{R}^m)$  is a metric space.

Hence (b) is proved.

(c) To prove that, if  $A \in L(\mathbb{R}^n, \mathbb{R}^m)$  and  $B \in L(\mathbb{R}^n, \mathbb{R}^m)$ ,

$$\text{then } \|BA\| = \|B\| \|A\|.$$

Consider 
$$\begin{aligned} |(BA)(x)| &= |B(Ax)| \\ &= \|B\| |Ax| \\ &= \|B\| \|A\| |x| \end{aligned}$$

Therefore,

$$\|BA\| = \sup\{|(BA)(x)|, x \in \mathbb{R}^n \text{ and } |x| = 1\} = \|B\| \|A\|$$

Hence  $\|BA\| = \|B\| \|A\|$

Hence (c) is proved.

Hence the theorem.

### Theorem 11.3

Let  $\mathcal{I}$  be the set of all invertible linear operators on  $\mathbb{R}^n$ .

(a) If  $A \in \mathcal{I}$ ,  $B \in L(\mathbb{R}^n)$  and  $\|B-A\| \|A^{-1}\| < 1$ , then  $B \in \mathcal{I}$ .

(b)  $\mathcal{I}$  is a subset of  $L(\mathbb{R}^n)$  and the mapping  $A \mapsto A^{-1}$  is continuous on  $\mathcal{I}$ .

[This mapping is also a one-to-one mapping of  $\mathcal{I}$  onto  $\mathcal{I}$ , which is its own inverse.]

#### Proof

(a) Let  $\|A^{-1}\| = 1/\alpha$  and  $\|B-A\| = \beta$ .

Then,

$$\|B-A\| \|A^{-1}\| < 1 \text{ implies } \beta/\alpha < 1$$

i.e.  $\beta < \alpha$ .

For every  $x \in \mathbb{R}^n$ ,

$$\begin{aligned} |x| &= |A^{-1}Ax| \\ &= \|A^{-1}\| |Ax| \\ &= \alpha |Ax| \end{aligned}$$

$$\begin{aligned}
 &= |Ax - Bx + Bx| \\
 &= |Ax - Bx| + |Bx| \\
 &= |(A - B)(x)| + |Bx| \\
 &= \|A - B\| |x| + \|Bx\| \\
 &= (\alpha - \beta) |x| + \|Bx\|,
 \end{aligned}$$

Therefore

$$(\alpha - \beta) |x| + \|Bx\| \leq \|Ax - Bx\|, \quad \text{for all } x \in \mathbb{R}^n. \quad \dots(1)$$

Since  $\alpha - \beta > 0$ ,

if  $x \neq 0$ , then  $Bx \neq 0$

Hence  $B$  is one-to-one.

Hence by theorem 11.1,  $R(B) = \mathbb{R}^n$ .

Hence  $B$  is onto.

Hence  $B$  is both one-to-one and onto.

Hence  $B$  is invertible,

Hence  $B \in GL(\mathbb{R}^n)$ .

(b) To prove that the mapping  $A \mapsto A^{-1}$  is continuous on  $GL(\mathbb{R}^n)$ .

Replacing  $x$  by  $B^{-1}y$  in (1), we get

$$(\alpha - \beta) \|B^{-1}y\| + \|B(B^{-1}y)\| = \|y\|, \quad \text{for all } y \in \mathbb{R}^n.$$

Therefore,

$$\|B^{-1}y\| \leq \frac{\|y\|}{(\alpha - \beta)}$$

Therefore,

$$\sup\{\|B^{-1}y\| / \|y\| \mid y \in \mathbb{R}^n \text{ and } \|y\| = 1\} = \frac{1}{(\alpha - \beta)}$$

Therefore,

$$\|B^{-1}\| = \frac{1}{(\alpha - \beta)}$$

Consider 
$$\begin{aligned}
 B^{-1} - A^{-1} &= B^{-1}A A^{-1} - B^{-1}B A^{-1} \\
 &= B^{-1}(A - B) A^{-1}
 \end{aligned}$$

Therefore,

$$\begin{aligned}
 \|B^{-1} - A^{-1}\| &= \|B^{-1}(A - B) A^{-1}\| \\
 &= \|B^{-1}\| \|A - B\| \|A^{-1}\| \\
 &= \frac{1}{(\alpha - \beta)} (\beta)
 \end{aligned}$$

This implies that  $\|B^{-1} - A^{-1}\| > 0$  as  $\beta > 0$  i.e. as  $\|B - A\| > 0$ .

Hence the mapping  $A \mapsto A^{-1}$  is continuous on  $GL(\mathbb{R}^n)$ .

Hence the theorem.

### 11.4 Matrix representation of a linear transformation

Suppose  $\{x_1, x_2, \dots, x_n\}$  and  $\{y_1, y_2, \dots, y_m\}$  are basis of the vector spaces  $X$  and  $Y$  respectively.

Let  $A \in L(X, Y)$ .

Since  $Ax_j \in Y$  for  $j=1, 2, \dots, n$ , and

$\{y_1, y_2, \dots, y_m\}$  is a basis of  $Y$ ,

$$Ax_j = \sum_{i=1}^m a_{ij} y_i, \quad j=1, 2, \dots, n. \quad \dots(1)$$

These numbers can be put into a rectangular array of  $m$  rows and  $n$  columns, called  $m$  by  $n$  matrix as follows:

$$[A] = \begin{bmatrix} a_{11} & a_{12} & \dots & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & \dots & a_{2n} \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ a_{m1} & a_{m2} & \dots & \dots & a_{mn} \end{bmatrix}$$

where the coordinates  $a_{ij}$  of the vectors  $Ax_j$  appear in the  $j$ th column of  $[A]$ . Therefore the vectors  $Ax_j$  are called column vectors of  $[A]$ .

With this terminology, the range of  $A$  is said to be spanned by the column vectors of  $[A]$ .

$$\begin{aligned} \text{If } x &= \sum_{j=1}^n c_j x_j, \\ Ax &= A\left(\sum_{j=1}^n c_j x_j\right) \\ &= \sum_{j=1}^n c_j Ax_j \\ &= \sum_{j=1}^n c_j \left(\sum_{i=1}^m a_{ij} y_i\right) \\ &= \sum_{i=1}^m \left(\sum_{j=1}^n a_{ij} c_j\right) y_i \end{aligned}$$

Therefore ,

$$Ax = \sum_{i=1}^m \left(\sum_{j=1}^n a_{ij} c_j\right) y_i \quad \dots(2).$$

Thus the coordinates of  $Ax$  are  $\sum_{j=1}^n a_{ij} c_j$ .

Suppose a  $m$  by  $n$  matrix is given with real entries  $a_{ij}$  and if  $A$  is defined as in (2), then  $A \in L(X, Y)$ . and  $[A]$  will be the given matrix.

Hence there is a one-to-one correspondence between  $L(X, Y)$  and the set of all  $m$  by  $n$  matrices.

Remark:

$[A]$  depends not only on  $A$  but also on the choice of bases in  $X$  and  $Y$ .

The same  $A$  may give rise to many different matrices, if we change bases and vice versa.

### 11.5 Let us sum up

In this lesson we have seen,

- Definition of vector spaces and some of its properties.
- Definition of linear transformation , sum of two linear transformations and product of two linear transformations

- Definion of norm of a linear transformation
- The set of all linear transformations from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  is a linear space
- The set of all linear transformations from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  is a metric space
- Definition of operators and inverse operators and
- How to represent a linear transformation in the matrix form.

### 11.6 Lesson End Activities

1. Write the matrix corresponding to the linear transformations  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  give by  $T(x, y, z) = (3x + z, -2x + y, x + 2y + z)$  with respect to the basis  $\{(1, 0, 1), (-1, 2, 1), (2, 1, 1)\}$  got both the domain & co-domain.

### 11.7 Reference

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## LESSON – 12

### CONTRACTION PRINCIPLE, INVERSE FUNCTION THEOREM AND IMPLICIT FUNCTION THEOREM.

#### 12.0 Introduction

#### 12.1 Aims and objectives

#### 12.2 Contraction Principle

#### 12.2 Inverse function theorem

#### 12.3 Implicit function theorem

#### 12.4 Let us sum up

#### 12.5 Lesson End Activities

#### 12.6 References

#### 12.0 Introduction

In this lesson we are going to study two important theorems in real analysis, inverse function theorem and implicit function theorem. The inverse function theorem states that a continuous differentiable mapping  $f$  is invertible in a neighborhood of any point  $x$  at which the linear transformation  $f'(x)$  is invertible. The implicit function theorem states that if  $f$  is continuously differentiable real function in the plane, then the equation  $f(x, y) = 0$  can be solved for  $y$  in terms of  $x$  in a neighborhood of any point  $(a, b)$  at which  $f(a, b) = 0$  and  $\partial f / \partial y \neq 0$ . Likewise, one can solve for  $x$  in terms of  $y$  near  $(a, b)$  if  $\partial f / \partial x \neq 0$ .

#### 12.1 Aims and objectives

After studying this lesson, you would know

- The definition of a contraction mapping
- Fixed point theorem
- Continuously differentiable mapping
- Inverse function theorem and
- Implicit function theorem and its linear version

#### 12.2 The contraction Principle.

##### Definition

Let  $X$  be a metric space with metric  $d$ .

If  $f$  maps  $X$  into  $X$  and if there is a number  $c < 1$  such that

$$d(f(x), f(y)) \leq cd(x, y), \quad \text{for all } x, y \in X,$$

then  $f$  is said to be a contraction of  $X$  into  $X$ .

##### Theorem 12.1

If  $X$  is a complete metric space and if  $f$  is a contraction of  $X$  into  $X$ , then there exists one and only one point  $x \in X$  such that

$$f(x) = x.$$

##### Proof

Choose  $x_0 \in X$  arbitrarily.

Define  $\{x_n\}$  recursively, by setting

$$x_{n+1} = f(x_n), \quad n=1,2,3,\dots \quad \dots(1)$$

Since  $f$  is a contraction of  $X$  into  $X$ ,

there exists a number  $c < 1$  such that

$$d(f(x), f(y)) \leq cd(x, y), \quad \text{for all } x, y \in X. \quad \dots(2)$$

Therefore,

$$d(x_{n+1}, x_n) = d(x_n, x_{n-1}) \cdot c$$

which in turn gives that

$$d(x_n, x_{n-1}) = c \cdot d(x_{n-1}, x_{n-2}).$$

Therefore,

$$d(x_{n+1}, x_n) = c^2 \cdot d(x_{n-1}, x_{n-2}).$$

Therefore, by induction,

$$d(x_{n+1}, x_n) = c^n \cdot d(x_1, x_0), \quad n=0,1,2,\dots$$

Therefore, if  $n < m$ ,

$$\begin{aligned} d(x_m, x_n) &= d(x_m, x_{m-1}) + d(x_{m-1}, x_{m-2}) + \dots + d(x_{n+2}, x_{n+1}) + d(x_{n+1}, x_n) \\ &= \sum_{i=n+1}^m d(x_i, x_{i-1}) \\ &= \sum_{i=n+1}^m c^{i-1} d(x_1, x_0) \\ &= (c^n + c^{n+1} + \dots + c^{m-1}) d(x_1, x_0) \\ &= c^n (1 + c + c^2 + \dots) d(x_1, x_0) \\ &= \frac{c^n}{1-c} d(x_1, x_0) \end{aligned}$$

Since  $c < 1$ ,

$$\lim_{n \rightarrow \infty} c^n = 0,$$

Therefore,

$$d(x_m, x_n) > 0 \text{ as } m, n \rightarrow \infty.$$

Hence  $\{x_n\}$  is a Cauchy sequence in  $X$ .

Since  $X$  is complete,

every Cauchy sequence of points of  $X$  converges to a point in  $X$ ,

Hence  $\{x_n\}$  converges to a point, say  $x$  in  $X$ .

i.e  $\lim_{n \rightarrow \infty} x_n = x$ .

Since  $f$  is a contraction,

$f$  is continuous on  $X$ .

Hence

$$\begin{aligned} f(x) &= f(\lim_{n \rightarrow \infty} x_n) \\ &= \lim_{n \rightarrow \infty} f(x_n) \\ &= \lim_{n \rightarrow \infty} x_{n+1} \\ &= x. \end{aligned}$$

Hence there exists a point  $x \in X$  such that  $f(x) = x$ .

Hence the theorem.

## 12.2 Inverse function theorem

### Definition

Suppose  $E$  is an open subset in  $\mathbb{R}^n$ ,  $f$  maps  $E$  into  $\mathbb{R}^m$  and  $x \in E$ .

$f$  is said to be differentiable at  $x$ ,

if there exists a linear transformation  $A$  of  $\mathbb{R}^n$  into  $\mathbb{R}^m$  such that

$$\lim_{h \rightarrow 0} \frac{|f(x+h) - f(x) - Ah|}{|h|} = 0$$

and we write  $f'(x) = A$ .

If  $f$  is differentiable at every  $x \in E$ , then  $f$  is said to be differentiable in  $E$ .

**Definition**

A differentiable mapping  $f$  of an open set  $E \subset \mathbb{R}^n$  into  $\mathbb{R}^m$  is said to be continuously differentiable in  $E$  if  $f'$  is a continuous mapping of  $E$  into  $L(\mathbb{R}^n, \mathbb{R}^m)$

i.e. if for every  $x \in E$  and for every  $\epsilon > 0$ ,

there exists a  $\delta > 0$  such that

$$\|f'(y) - f'(x)\| < \epsilon, \quad \text{if } y \in E \text{ and } |x-y| < \delta,$$

then  $f$  is said to be  $C^1$  mapping or  $f \in C^1(E)$ .

**Theorem 12.2 Inverse function theorem**

Suppose  $f$  is a  $C^1$ -mapping of an open subset  $E \subset \mathbb{R}^n$  into  $\mathbb{R}^n$ ,  $f'(a)$  is invertible for some  $a \in E$  and  $b=f'(a)$ . Then

(a) there exist open sets  $U$  and  $V$  in  $\mathbb{R}^n$  such that  $a \in U, b \in V$ ,  $f$  is one-to-one on  $U$  and  $f(U) = V$ .

(b) If  $g$  is the inverse of  $f$ , defined in  $V$  by

$$g(f(x))=x, \quad x \in U,$$

then  $g \in C^1(V)$ .

[ Writing the equation  $y=f(x)$  in component form, the conclusion of the theorem can be interpreted as follows:

The system of  $n$  equations

$$y_i = f_i(x_1, x_2, \dots, x_n) \quad i=1,2,\dots,n$$

can be solved for  $x_1, x_2, \dots, x_n$  in terms of  $y_1, y_2, \dots, y_n$ .

If we restrict  $x$  and  $y$  to be small enough neighborhoods of  $a$  and  $b$ , the solutions are unique and continuously differentiable. ]

**Proof**

(a) Let  $A = f'(a)$ .

Choose  $\delta$  such that

$$\|A^{-1}\| = 1 \tag{1}$$

Since  $f$  is a  $C^1$ -mapping on  $E$  and  $a \in E$ ,

$f'$  is continuous at  $a$ .

Hence there exists an open ball  $U \subset E$ , with centre at  $a$  such that

$$\|f'(x) - A\| < \delta, \quad \text{for all } x \in U. \tag{2}$$

To each  $y \in \mathbb{R}^n$ , we associate a function  $\phi$  on  $E$  as

$$\phi(x) = x + A^{-1}(y - f(x)), \quad x \in E \tag{3}$$

so that if  $f(x)=y$ , then  $\phi(x) = x$ .

Conversely,

$$\text{if } \phi(x) = x,$$

then

$$A^{-1}(y - f(x)) = 0$$

implies

$$y - f(x) = 0$$

implies

$$y = f(x)$$

Hence  $f(x) = y$  if and only if  $\phi(x) = x$

i.e.  $x$  is a fixed point of  $\phi$ . ... (4)

Now

$$\begin{aligned} \phi'(x) &= I - A^{-1}f'(x) \\ &= A^{-1}A - A^{-1}f'(x) \\ &= A^{-1}(A - f'(x)) \end{aligned}$$

Therefore for  $\mathbf{x} \in U$ ,

$$\begin{aligned} \|\mathbf{f}'(\mathbf{x})\| &= \|A^{-1}(A - \mathbf{f}''(\mathbf{x}))\| \\ &= \|A^{-1}\| \|A - \mathbf{f}''(\mathbf{x})\| \\ &\leq (1/2) \quad [\text{from (1) and (2)}] \end{aligned}$$

Therefore,

$$\|\mathbf{f}'(\mathbf{x})\| \leq 1/2 \quad \text{for all } \mathbf{x} \in U.$$

Therefore by the result,

“Suppose  $\mathbf{f}$  maps a convex open set  $E \subset \mathbb{R}^n$  into  $\mathbb{R}^m$ .

If  $\mathbf{f}$  is differentiable in  $E$ , and there is a real number  $M$  such that

$$\|\mathbf{f}''(\mathbf{x})\| \leq M \quad \text{for every } \mathbf{x} \in E.$$

Then  $\|\mathbf{f}(\mathbf{b}) - \mathbf{f}(\mathbf{a}) - \mathbf{f}'(\mathbf{a})(\mathbf{b} - \mathbf{a})\| \leq M \|\mathbf{b} - \mathbf{a}\|^2$  for all  $\mathbf{a} \in E, \mathbf{b} \in E$ .”,

we get that ,

$$\|\mathbf{f}(\mathbf{x}_1) - \mathbf{f}(\mathbf{x}_2) - \mathbf{f}'(\mathbf{x}_0)(\mathbf{x}_1 - \mathbf{x}_2)\| \leq (1/2)M \|\mathbf{x}_1 - \mathbf{x}_2\|^2, \quad \text{for all } \mathbf{x}_1, \mathbf{x}_2 \in U. \quad \dots(5)$$

Therefore,

$\mathbf{f}$  is a contraction of  $U$  into  $U$ .

Therefore ,

$\mathbf{f}$  has at most one fixed point in  $U$ . [by theorem 12.1]

Therefore ,

$\mathbf{f}(\mathbf{x}) = \mathbf{y}$  for at most one  $\mathbf{x} \in U$ . [From (4)]

Hence  $\mathbf{f}$  is one-to-one in  $U$ .

Let  $V = \mathbf{f}(U)$ .

To prove that  $V$  is open.

Let  $\mathbf{y}_0 \in V$ .

Then  $\mathbf{y}_0 = \mathbf{f}(\mathbf{x}_0)$  for some  $\mathbf{x}_0 \in U$ .

Let  $B$  be the open ball with centre at  $\mathbf{x}_0$  and radius  $r > 0$ , so small that its closure  $\bar{B}$  lies in  $U$ .

Claim:  $\mathbf{y} \in V$  whenever  $\|\mathbf{y} - \mathbf{y}_0\| < r$ .

Let  $\mathbf{y}$  be such that  $\|\mathbf{y} - \mathbf{y}_0\| < r$ .

$$\begin{aligned} \text{From (3), } \|\mathbf{f}(\mathbf{x}_0) - \mathbf{y}\| &= \|A^{-1}(\mathbf{y} - \mathbf{y}_0)\| \\ &= \|A^{-1}\| \|\mathbf{y} - \mathbf{y}_0\| \\ &< \|A^{-1}\| r \\ &= (1/2)r \quad [\text{from (1)}] \end{aligned}$$

Hence,

$$\|\mathbf{f}(\mathbf{x}_0) - \mathbf{y}\| < r/2. \quad \dots(6)$$

If  $\mathbf{x} \in B$ , then by (5),

$$\|\mathbf{f}(\mathbf{x}) - \mathbf{f}(\mathbf{x}_0) - \mathbf{f}'(\mathbf{x}_0)(\mathbf{x} - \mathbf{x}_0)\| \leq (1/2)M \|\mathbf{x} - \mathbf{x}_0\|^2 \quad \dots(7)$$

From (6) and (7), we get that

whenever  $\mathbf{x} \in B$ ,

$$\begin{aligned} \|\mathbf{f}(\mathbf{x}) - \mathbf{y}\| &= \|\mathbf{f}(\mathbf{x}) - \mathbf{f}(\mathbf{x}_0) + \mathbf{f}(\mathbf{x}_0) - \mathbf{y}\| \\ &\leq \|\mathbf{f}(\mathbf{x}) - \mathbf{f}(\mathbf{x}_0) - \mathbf{f}'(\mathbf{x}_0)(\mathbf{x} - \mathbf{x}_0)\| + \|\mathbf{f}(\mathbf{x}_0) - \mathbf{y}\| \\ &< (1/2)M \|\mathbf{x} - \mathbf{x}_0\|^2 + (r/2) \\ &= (r/2) + (r/2) = r. \end{aligned}$$

Hence ,

$$\|\mathbf{f}(\mathbf{x}) - \mathbf{y}\| < r \quad \text{whenever } \mathbf{x} \in B.$$

This implies that  $\mathbf{f}(\mathbf{x}) \in V$ .

Hence  $\mathbf{f}$  maps  $B$  into  $V$ .

Since  $\mathbf{f}$  is a contraction of  $U$  into  $U$  and  $\bar{B} \subset U$ ,

is a contraction of  $\bar{B}$  into  $\bar{B}$ .

Since  $\bar{B}$  is a closed subset of  $\mathbb{R}^n$  and  $\mathbb{R}^n$  is complete,  $\bar{B}$  is complete.

Hence by the theorem 12.1,

has a fixed point  $\mathbf{x} \in \bar{B}$ .

i.e.  $\mathbf{f}(\mathbf{x}) = \mathbf{x}$ .

For this  $\mathbf{x}$ ,  $\mathbf{f}(\mathbf{x}) = \mathbf{y}$ .

[From (4)]

Hence,

$$\mathbf{y} \in \mathbf{f}(\bar{B}) \subset \mathbf{f}(U) = V.$$

Hence,

$$\mathbf{y} \in V \text{ whenever } |\mathbf{y} - \mathbf{y}_0| < r.$$

Hence  $V$  is open.

Hence part(a) of the theorem is proved.

To prove part(b):

Let  $\mathbf{y} \in V$ ,  $\mathbf{y} + \mathbf{t} \in V$ .

Since  $V = \mathbf{f}(U)$ ,

there exists a  $\mathbf{x} \in U$ ,  $\mathbf{x} + \mathbf{h} \in U$  such that

$$\mathbf{y} = \mathbf{f}(\mathbf{x}) \text{ and}$$

$$\mathbf{y} + \mathbf{k} = \mathbf{f}(\mathbf{x} + \mathbf{h}).$$

From (3),

$$\begin{aligned} (\mathbf{x} + \mathbf{h}) - \mathbf{x} &= [(\mathbf{x} + \mathbf{h}) + A^{-1}(\mathbf{y} - \mathbf{f}(\mathbf{x} + \mathbf{h}))] - [\mathbf{x} + A^{-1}(\mathbf{y} - \mathbf{f}(\mathbf{x}))] \\ &= \mathbf{h} + A^{-1}(\mathbf{f}(\mathbf{x}) - \mathbf{f}(\mathbf{x} + \mathbf{h})) \\ &= \mathbf{h} - A^{-1}\mathbf{k} \end{aligned} \quad \dots(8)$$

By(5),

$$\begin{aligned} |(\mathbf{x} + \mathbf{h}) - \mathbf{x}| &\leq (1/2) |(\mathbf{x} + \mathbf{h}) - \mathbf{x}| \\ &= (1/2)|\mathbf{h}|, \end{aligned} \quad \dots(9)$$

From (8) and (9), we get

$$|\mathbf{h} - A^{-1}\mathbf{k}| = |(\mathbf{x} + \mathbf{h}) - \mathbf{x}| \leq (1/2)|\mathbf{h}|$$

Therefore,

$$\begin{aligned} |A^{-1}\mathbf{k}| &= |\mathbf{h} - (\mathbf{h} - A^{-1}\mathbf{k})| \\ &= |\mathbf{h}| - |\mathbf{h} - A^{-1}\mathbf{k}| \\ &= |\mathbf{h}| - |\mathbf{h}|/2 = |\mathbf{h}|/2 \end{aligned}$$

Hence

$$|A^{-1}\mathbf{k}| \leq |\mathbf{h}|/2$$

Therefore,

$$\begin{aligned} |\mathbf{h}| &\leq 2|A^{-1}\mathbf{k}| \\ &\leq 2\|A^{-1}\|\|\mathbf{k}\| \\ &= (1/\lambda)\|\mathbf{k}\| \end{aligned} \quad \dots(10)$$

Therefore,

$$\frac{1}{\|\mathbf{k}\|} \leq \frac{1}{\lambda\|\mathbf{h}\|} \quad \dots(11)$$

Let  $\mathbf{g}$  be the inverse of  $\mathbf{f}$ , defined in  $V$  by

$$\mathbf{g}(\mathbf{f}(\mathbf{x})) = \mathbf{x}, \quad \mathbf{x} \in U.$$

Let  $\mathbf{x} \in U$ .

From (2),

$$\|\mathbf{f}'(\mathbf{x}) - A\| < \epsilon = 1/\|A^{-1}\|$$

i.e.  $\|\mathbf{f}'(\mathbf{x}) - A\| < 1/\|A^{-1}\|$ , where  $A = \mathbf{f}'(\mathbf{a})$  is invertible.

Hence by theorem 11.3

$f'(x)$  is invertible.

Let the inverse of  $f'(x)$  be  $T$ .

$$\begin{aligned} \text{Consider } g(y+k) - g(y) - Tk &= g(f(x+h)) - g(f(x)) - Tk \\ &= f(x+h) - f(x) - Tk \\ &= h - Tk \\ &= Tf'(x)h - Tk \\ &\quad [\text{since } T \text{ is the inverse of } f'(x), Tf'(x)=I] \\ &= -T [k - f'(x)h] \\ &= -T [f(x+h) - f(x) - f'(x)h] \end{aligned}$$

Therefore,

$$|g(y+k) - g(y) - Tk| = |T [f(x+h) - f(x) - f'(x)h]| \tag{12}$$

From (11) and (12),

$$\frac{|g(y+k) - g(y) - Tk|}{|k|} = \frac{\|T\| |f(x+h) - f(x) - f'(x)h|}{\lambda |h|} \tag{13}$$

From (10), it is clear that  $h \rightarrow 0$  as  $k \rightarrow 0$ .

Also, since  $f$  is differentiable at  $x$ ,

$$\lim_{h \rightarrow 0} \frac{|f(x+h) - f(x) - f'(x)h|}{|h|} = 0.$$

Therefore from (13),

$$\lim_{k \rightarrow 0} \frac{|g(y+k) - g(y) - Tk|}{|k|} = 0.$$

This implies that

$$\begin{aligned} g'(y) \text{ exists and} \\ g'(y) = T. \end{aligned}$$

Since  $T$  is the inverse of  $f'(x) = f'(g(y))$ ,

$$g'(y) = \{f'(g(y))\}^{-1} \text{ for all } y \in V.$$

Hence  $g$  is differentiable on  $V$ .

Hence  $g$  is a continuous mapping of  $V$  onto  $U$ .

Also  $f'$  is a continuous mapping  $U$  into the set of all invertible elements of  $L(\mathbb{R}^n)$ .

Also the inversion is a continuous mapping onto .

Hence  $g'$  is a continuous mapping on  $V$ .

Hence  $g \in C^1(V)$ .

Hence the theorem.

### Theorem 12.3

If  $f$  is a  $C^1$ - mapping of an open set  $E \subset \mathbb{R}^n$  into  $\mathbb{R}^m$  and

if  $f'(x)$  is invertible for every  $x \in E$ ,

then  $f(W)$  is an open subset of  $\mathbb{R}^m$  for every open set  $W \subset E$ .

In other words,  $f$  is an open mapping of  $E$  into  $\mathbb{R}^m$ .

#### Proof

Let  $W \subset E$  is an open set.

To prove that  $f(W)$  is an open subset of  $\mathbb{R}^m$ .

Let  $b \in f(W)$ .

Then there exists a  $a \in W$  such that  $b = f(a)$ .

Since  $f'(x)$  is invertible for every  $x \in E$ , and  $a \in W \subset E$ ,

$f'(a)$  is invertible.

Hence by part (a) of Theorem 12.2,

there exist open sets  $U_a \subset W$  and  $V_b \subset f(W)$  such that  $\mathbf{a} \in U$ ,  $\mathbf{b} \in V$ ,  $f$  is one-to-one on  $U$  and  $f(U) = V$ .

Hence  $f(W)$  can be written as  $f(W) = \bigcup_{b \in f(W)} V_b$ .

Since union of open sets is open,  $f(W)$  is open.

Hence the theorem.

### 12.3 Implicit function theorem

#### Notation

If  $\mathbf{x} = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$  and  
 $\mathbf{y} = (y_1, y_2, \dots, y_m) \in \mathbb{R}^m$ ,

We write,

$(\mathbf{x}, \mathbf{y}) = (x_1, x_2, \dots, x_n, y_1, y_2, \dots, y_m) \in \mathbb{R}^{n+m}$ .

The first entry in  $(\mathbf{x}, \mathbf{y})$  or in a similar symbol will always be a vector in  $\mathbb{R}^n$ , the second will be a vector in  $\mathbb{R}^m$ .

Every vector  $A \in L(\mathbb{R}^{n+m}, \mathbb{R}^n)$  can be split into two linear transformation  $A_x$  and  $A_y$ , defined by

$$\begin{aligned} A_x \mathbf{h} &= A(\mathbf{h}, \mathbf{0}), \\ A_y \mathbf{k} &= A(\mathbf{0}, \mathbf{k}), \text{ for any } \mathbf{h} \in \mathbb{R}^n, \mathbf{k} \in \mathbb{R}^m. \end{aligned}$$

Then  $A_x \in L(\mathbb{R}^n, \mathbb{R}^n)$ ,  $A_y \in L(\mathbb{R}^m, \mathbb{R}^n)$  and

$$\begin{aligned} A(\mathbf{h}, \mathbf{k}) &= A[(\mathbf{h}, \mathbf{0}) + (\mathbf{0}, \mathbf{k})] \\ &= A(\mathbf{h}, \mathbf{0}) + A(\mathbf{0}, \mathbf{k}) \quad [\text{since } A \text{ is linear}] \\ &= A_x \mathbf{h} + A_y \mathbf{k}. \end{aligned}$$

#### Theorem 12.4 [Linear version of the implicit function theorem]

If  $A \in L(\mathbb{R}^{n+m}, \mathbb{R}^n)$  and if  $A_x$  is invertible, then there corresponds to every  $\mathbf{k} \in \mathbb{R}^m$  a unique  $\mathbf{h} \in \mathbb{R}^n$  such that  $A(\mathbf{h}, \mathbf{k}) = \mathbf{0}$ . This  $\mathbf{h}$  can be computed from  $\mathbf{k}$  by the formula

$$\mathbf{h} = -(A_x)^{-1} A_y \mathbf{k}.$$

#### Proof

Now  $A(\mathbf{h}, \mathbf{k}) = \mathbf{0}$  if and only if  $A_x \mathbf{h} + A_y \mathbf{k} = \mathbf{0}$ .

Since  $A_x$  is invertible,

$(A_x)^{-1}$  exists.

Therefore premultiplying  $A_x \mathbf{h} + A_y \mathbf{k} = \mathbf{0}$  by  $(A_x)^{-1}$ , we get

$$\begin{aligned} (A_x)^{-1}(A_x \mathbf{h} + A_y \mathbf{k}) &= \mathbf{0} \\ \mathbf{h} + (A_x)^{-1} A_y \mathbf{k} &= \mathbf{0}. \end{aligned}$$

Therefore,

$$\mathbf{h} = -(A_x)^{-1} A_y \mathbf{k}.$$

Hence the theorem.

#### Check your progress

Prove that  $A_x$  and  $A_y$  are linear.

#### Theorem 12.5 The implicit function theorem

Let  $f$  be a  $C^1$ -mapping of an open set  $E \subset \mathbb{R}^{n+m}$  into  $\mathbb{R}^n$ , such that

$f(\mathbf{a}, \mathbf{b}) = \mathbf{0}$  for some point  $(\mathbf{a}, \mathbf{b}) \in E$ .

Put  $A = f'(\mathbf{a}, \mathbf{b})$  and assume that  $A_x$  is invertible.

Then there exist open sets  $U \subset \mathbb{R}^{n+m}$  and  $W \subset \mathbb{R}^m$ , with  $(\mathbf{a}, \mathbf{b}) \in U$  and  $\mathbf{b} \in W$ , having the following property:

To every  $\mathbf{y} \in W$  corresponds a unique  $\mathbf{x}$  such that

$$(\mathbf{x}, \mathbf{y}) \in U \quad \text{and} \quad \mathbf{f}(\mathbf{x}, \mathbf{y}) = \mathbf{0}. \quad \dots(1)$$

If this  $\mathbf{x}$  is defined to be  $\mathbf{g}(\mathbf{y})$ ,

then  $\mathbf{g}$  is a  $C^1$ -mapping of  $W$  into  $\mathbb{R}^n$ ,

$$\begin{aligned} \mathbf{g}(\mathbf{b}) &= \mathbf{a}, \\ \mathbf{f}(\mathbf{g}(\mathbf{y}), \mathbf{y}) &= \mathbf{0} \quad \mathbf{y} \in W. \end{aligned} \quad \dots(2)$$

and

$$\mathbf{g}'(\mathbf{b}) = - (A_x)^{-1} A_y. \quad \dots(3)$$

The function  $\mathbf{g}$  is implicitly defined by (2).

Hence the name of the theorem.

**Proof**

Define  $\mathbf{F}: E \rightarrow \mathbb{R}^{n+m}$  by

$$\mathbf{F}(\mathbf{x}, \mathbf{y}) = (\mathbf{f}(\mathbf{x}, \mathbf{y}), \mathbf{y}) \quad \text{for} \quad (\mathbf{x}, \mathbf{y}) \in E.$$

Since  $\mathbf{f}$  is a  $C^1$ -mapping,

$\mathbf{F}$  is a  $C^1$ -mapping of  $E$  into  $\mathbb{R}^{n+m}$ .

Claim:  $\mathbf{F}'(\mathbf{a}, \mathbf{b})$  is an invertible element of  $L(\mathbb{R}^{n+m})$

Since  $\mathbf{f}'(\mathbf{a}, \mathbf{b})$  exists and equals  $A$ ,

$$\mathbf{f}(\mathbf{a}+\mathbf{h}, \mathbf{b}+\mathbf{k}) = \mathbf{f}(\mathbf{a}, \mathbf{b}) + A(\mathbf{h}, \mathbf{k}) + \mathbf{r}(\mathbf{h}, \mathbf{k}),$$

where  $\mathbf{r}$  is the remainder that occurs in the definition of  $\mathbf{f}'(\mathbf{a}, \mathbf{b})$ .

Therefore,

$$\mathbf{f}(\mathbf{a}+\mathbf{h}, \mathbf{b}+\mathbf{k}) - \mathbf{f}(\mathbf{a}, \mathbf{b}) = A(\mathbf{h}, \mathbf{k}) + \mathbf{r}(\mathbf{h}, \mathbf{k}). \quad [\text{Since } \mathbf{f}(\mathbf{a}, \mathbf{b}) = \mathbf{0}] \quad \dots(4)$$

Consider  $\mathbf{F}(\mathbf{a}+\mathbf{h}, \mathbf{b}+\mathbf{k}) - \mathbf{F}(\mathbf{a}, \mathbf{b}) = (\mathbf{f}(\mathbf{a}+\mathbf{h}, \mathbf{b}+\mathbf{k}), \mathbf{b}+\mathbf{k}) - (\mathbf{f}(\mathbf{a}, \mathbf{b}), \mathbf{b})$

$$= (\mathbf{f}(\mathbf{a}+\mathbf{h}, \mathbf{b}+\mathbf{k}), \mathbf{b}+\mathbf{k}) - (\mathbf{0}, \mathbf{b})$$

$$= (\mathbf{f}(\mathbf{a}+\mathbf{h}, \mathbf{b}+\mathbf{k}), \mathbf{k})$$

$$= (A(\mathbf{h}, \mathbf{k}) + \mathbf{r}(\mathbf{h}, \mathbf{k}), \mathbf{k} + \mathbf{0}) \quad [\text{from (4)}]$$

$$= (A(\mathbf{h}, \mathbf{k}), \mathbf{k}) + (\mathbf{r}(\mathbf{h}, \mathbf{k}), \mathbf{0})$$

Therefore,

$\mathbf{F}'(\mathbf{a}, \mathbf{b})$  is the linear operator on  $\mathbb{R}^{n+m}$  that maps  $(\mathbf{h}, \mathbf{k})$  to  $(A(\mathbf{h}, \mathbf{k}), \mathbf{k})$ .

If this image vector is  $\mathbf{0}$ ,

$$\text{then } A(\mathbf{h}, \mathbf{k}) = \mathbf{0} \text{ and } \mathbf{k} = \mathbf{0}.$$

Hence  $A(\mathbf{h}, \mathbf{0}) = \mathbf{0}$ .

This implies  $\mathbf{h} = \mathbf{0}$ . [By theorem 12.4]

Therefore if the image vector of  $\mathbf{F}'(\mathbf{a}, \mathbf{b})$  at  $(\mathbf{h}, \mathbf{k})$  is  $\mathbf{0}$ ,

$$\text{then both } \mathbf{h} = \mathbf{0} \text{ and } \mathbf{k} = \mathbf{0}.$$

Hence  $\mathbf{F}'(\mathbf{a}, \mathbf{b})$  is one-to-one.

Hence by theorem 11.1,

$$\mathbf{F}'(\mathbf{a}, \mathbf{b}) \text{ is onto.}$$

Hence  $\mathbf{F}'(\mathbf{a}, \mathbf{b})$  is both one-to-one and onto.

Hence  $\mathbf{F}'(\mathbf{a}, \mathbf{b})$  is invertible.

Hence our claim is proved.

Hence inversion theorem can be applied to  $\mathbf{F}$ .

Hence by part(a) of the inversion theorem,

there exist open sets  $U$  and  $V$  in  $\mathbb{R}^{n+m}$  with  $(\mathbf{a}, \mathbf{b}) \in U$ ,  $(\mathbf{0}, \mathbf{b}) \in V$ , such that

$\mathbf{F}$  is a one-to-one mapping of  $U$  onto  $V$ .

Let  $W = \{\mathbf{y} \in \mathbb{R}^m / (\mathbf{0}, \mathbf{y}) \in V\}$

Therefore  $\mathbf{b} \in W$ .

[since  $(\mathbf{0}, \mathbf{b}) \in V$ ].

Since  $V$  is open,

$W$  is also open.

If  $\mathbf{y} \in W$ ,

then  $(\mathbf{0}, \mathbf{y}) \in V = F(U)$ .

Hence there exists a point  $(\mathbf{x}, \mathbf{y})$  in  $R^{n+m}$  such that

$$(\mathbf{0}, \mathbf{y}) = F(\mathbf{x}, \mathbf{y}).$$

Therefore ,

$$(\mathbf{0}, \mathbf{y}) = (f(\mathbf{x}, \mathbf{y}), \mathbf{y})$$

This implies that

$$f(\mathbf{x}, \mathbf{y}) = \mathbf{0}, \text{ for this } \mathbf{x}.$$

Suppose for the same  $\mathbf{y}$ , there exists another  $\mathbf{x}'$  in  $R^n$ , such that

$$f(\mathbf{x}', \mathbf{y}) = \mathbf{0}.$$

Then

$$\begin{aligned} F(\mathbf{x}', \mathbf{y}) &= (f(\mathbf{x}', \mathbf{y}), \mathbf{y}) \\ &= (\mathbf{0}, \mathbf{y}) \\ &= (f(\mathbf{x}, \mathbf{y}), \mathbf{y}) \\ &= F(\mathbf{x}, \mathbf{y}). \end{aligned}$$

Since  $F$  is one-to-one in  $U$ ,

this implies that  $\mathbf{x}' = \mathbf{x}$ .

Hence this  $\mathbf{x}$  is unique.

Hence there exist open sets  $U \subset R^{n+m}$  and  $W \subset R^m$ , with  $(\mathbf{a}, \mathbf{b}) \in U$  and  $\mathbf{b} \in W$ , with the property that:

To every  $\mathbf{y} \in W$  corresponds a unique  $\mathbf{x}$  such that

$$\begin{aligned} (\mathbf{x}, \mathbf{y}) &\in U \quad \text{and} \\ f(\mathbf{x}, \mathbf{y}) &= \mathbf{0}. \end{aligned}$$

Hence first part of the theorem is proved.

To prove the second part:

As we have seen in the first part,

to every  $\mathbf{y} \in W$  there is a unique  $\mathbf{x}$  such that

$$(\mathbf{x}, \mathbf{y}) \in U \quad \text{and} \quad f(\mathbf{x}, \mathbf{y}) = \mathbf{0}.$$

Define this  $\mathbf{x}$  to be  $\mathbf{g}(\mathbf{y})$  for every  $\mathbf{y} \in W$ .

Then  $\mathbf{g}(\mathbf{b}) = \mathbf{a}$ . [since  $f(\mathbf{a}, \mathbf{b}) = \mathbf{0}$ ]

Also  $(\mathbf{g}(\mathbf{y}), \mathbf{y}) = (\mathbf{x}, \mathbf{y}) \in U$  and

$$f(\mathbf{g}(\mathbf{y}), \mathbf{y}) = \mathbf{0}. \quad \dots(5)$$

Therefore ,

$$F(\mathbf{g}(\mathbf{y}), \mathbf{y}) = (f(\mathbf{g}(\mathbf{y}), \mathbf{y}), \mathbf{y}) = (\mathbf{0}, \mathbf{y}), \text{ for all } \mathbf{y} \in W. \quad \dots(6)$$

Let  $\mathbf{G}$  be the mapping of  $V$  into  $U$  that inverts  $F$  i.e  $\mathbf{G} = F^{-1}$ .

Also  $F \in C'$ , implies  $\mathbf{G} \in C'$

Then from(6),

$$\begin{aligned} \mathbf{G}[F(\mathbf{g}(\mathbf{y}), \mathbf{y})] &= \mathbf{G}(\mathbf{0}, \mathbf{y}), \text{ for all } \mathbf{y} \in W. \\ \text{i.e } (\mathbf{g}(\mathbf{y}), \mathbf{y}) &= \mathbf{G}(\mathbf{0}, \mathbf{y}), \text{ for all } \mathbf{y} \in W. \end{aligned} \quad \dots(7)$$

Since  $\mathbf{G} \in C'$ , (7) shows that  $\mathbf{g} \in C'$ .

To compute  $\mathbf{g}'(\mathbf{b})$ :

Let  $(\mathbf{g}(\mathbf{y}), \mathbf{y}) = (\mathbf{y})$ .

Then  $\mathbf{y}'\mathbf{k} = (\mathbf{g}'(\mathbf{y})\mathbf{k}, \mathbf{y}\mathbf{k})$  for all  $\mathbf{y} \in W$  and  $\mathbf{k} \in R^m$ . ... (8)

From (5),

$$f(\mathbf{y}) = f(\mathbf{g}(\mathbf{y}), \mathbf{y}) = \mathbf{0}.$$

Therefore by chain rule,



Take  $\mathbf{a}=(0, 1)$  and  $\mathbf{b}=(3,2,7)$ ,  
then  $\mathbf{f}(\mathbf{a}, \mathbf{b})=0$ .

Now  $D_1f_1(x_1, x_2, y_1, y_2, y_3) = 2e^{-x_1}$

Therefore ,

$$D_1f_1(0, 1, 3, 2, 7) = 2$$

Similarly ,

$$D_2f_1(0, 1, 3, 2, 7) = 3$$

$$D_3f_1(0, 1, 3, 2, 7) = 1$$

$$D_4f_1(0, 1, 3, 2, 7) = -4$$

$$D_5f_1(0, 1, 3, 2, 7) = 0$$

$$D_1f_2(0, 1, 3, 2, 7) = -6$$

$$D_2f_2(0, 1, 3, 2, 7) = 1$$

$$D_3f_2(0, 1, 3, 2, 7) = 2$$

$$D_4f_2(0, 1, 3, 2, 7) = 0$$

$$D_5f_2(0, 1, 3, 2, 7) = -1.$$

Therefore the matrix of the transformation  $A=\mathbf{f}'(\mathbf{a}, \mathbf{b})$  is

$$[A] = \begin{bmatrix} 2 & 3 & 1 & -4 & 0 \\ -6 & 1 & 2 & 0 & -1 \end{bmatrix}.$$

Hence,

$$[A_x] = \begin{bmatrix} 2 & 3 \\ -6 & 1 \end{bmatrix},$$

And  $[A_y] = \begin{bmatrix} 1 & -4 & 0 \\ 2 & 0 & -1 \end{bmatrix}.$

Since the column vectors of  $[A_x]$  i.e  $\begin{bmatrix} 2 \\ -6 \end{bmatrix}$  and  $\begin{bmatrix} 3 \\ 1 \end{bmatrix}$  are independent,

$A_x$  is invertible.

Therefore by the implicit function theorem ,

a  $C^1$ -mapping  $\mathbf{g}$  exists in a neighborhood of  $\mathbf{b}=(3,2,7)$  such that

$$\mathbf{g}(\mathbf{b}) = \mathbf{g}(3, 2, 7)=(0, 1)=\mathbf{a}$$

$$\text{and } \mathbf{f}(\mathbf{g}(\mathbf{y}), \mathbf{y})=0.$$

$$[(A_x)^{-1}] = \frac{1}{20} \begin{bmatrix} 1 & -3 \\ 6 & 2 \end{bmatrix}$$

Moreover,

from  $\mathbf{g}'(\mathbf{b}) = -(A_x)^{-1}A_y$ , we get

$$\begin{aligned} \mathbf{g}'(3, 2, 7) &= -\frac{1}{20} \begin{bmatrix} 1 & -3 \\ 6 & 2 \end{bmatrix} \begin{bmatrix} 1 & -4 & 0 \\ 2 & 0 & -1 \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{4} & \frac{1}{5} & -\frac{3}{20} \\ -\frac{1}{2} & \frac{6}{5} & \frac{1}{10} \end{bmatrix}. \end{aligned}$$

In terms of partial derivatives,

the conclusion is:

At the point  $(3, 2, 7)$ ,

$$D_1g_1 = 1/4 \quad D_2g_1 = 1/5 \quad D_3g_1 = -3/20$$

$$D_1g_2 = -1/2 \quad D_2g_2 = 6/5 \quad D_3g_2 = 1/10.$$

### 12.4 Let us sum up

In this lesson we have studied

- Contraction mapping
- Fixed point theorem
- Inverse function theorem
- Implicit function theorem and
- To compute the derivative of the inverse function at a given point.

### 12.5 Lesson End Activities

Let  $f = (f_1, f_2)$  be the mapping of  $\mathbb{R}^2$  into  $\mathbb{R}^2$  given by

$$f_1(x, y) = e^x \cos y,$$

$$f_2(x, y) = e^x \sin y.$$

Put  $\mathbf{a} = (0, \pi/3)$ ,  $\mathbf{b} = f(\mathbf{a})$ .

1. Find an explicit formula for  $g$ .
2. Compute  $f'(\mathbf{a})$  and  $g'(\mathbf{b})$
3. Verify the formula  $g'(\mathbf{y}) = \{f'(g(\mathbf{y}))\}^{-1}$  when  $\mathbf{y} = \mathbf{b}$ .

### 12.6 References

1. Principles of Mathematical Analysis by Walter Rudin.

## Lesson 13 Determinants

### 13.0 Introduction

#### 13.1 Aims and objectives

#### 13.2 Determinants

#### 13.3 Jacobians

#### 13.4 Let us sum up

#### 13.5 Lesson End Activities

#### 13.6 References

### 13.0 Introduction

Determinants are numbers associated to square matrices, and hence to the operators represented by such matrices. In this lesson we are going to study the definition of determinants and some of its properties.

#### 13.1 Aims and objectives

After studying this lesson, you would know

- The general definition of the determinant
- Important properties of the determinant
- Jacobians

#### 13.2 Determinants

##### Definition

If  $(j_1, j_2, \dots, j_n)$  is an ordered  $n$ -tuple of integers, the function  $s(j_1, j_2, \dots, j_n)$  is defined as

$$s(j_1, j_2, \dots, j_n) = \prod_{p < q} \text{sgn}(j_q - j_p),$$

where

$$\begin{aligned} \text{sgn } x &= 1 && \text{if } x > 0 \\ \text{sgn } x &= -1 && \text{if } x < 0 \\ \text{sgn } x &= 0 && \text{if } x = 0. \end{aligned}$$

Thus  $s(j_1, j_2, \dots, j_n) = 1, -1$ , or  $0$ ,

and it changes sign if any two of the  $j$ 's are interchanged.

##### Definition

Let  $[A]$  be the matrix of a linear operator  $A$  on  $\mathbb{R}^n$ , relative to the standard basis  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$ , with entries  $a(i, j)$  in the  $i$ th row and  $j$ th column.

The determinant of  $[A]$  is defined to be the number

$$\det [A] = s(j_1, j_2, \dots, j_n) a(1, j_1) a(2, j_2) \dots a(n, j_n).$$

The sum extends over all ordered  $n$ -tuples of integers  $(j_1, j_2, \dots, j_n)$  with  $1 \leq j \leq n$ .

The column vectors  $\mathbf{x}_j$  of  $[A]$  are

$$\mathbf{x}_j = \sum_{i=1}^n a(i, j) \mathbf{e}_i \quad (1 \leq j \leq n)$$

It will be convenient to think of  $\det [A]$  as a function of the column vectors of  $[A]$ .

If we write

$$\det [A] = \det (\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n),$$

then  $\det$  is now a real function on the set of all ordered  $n$ -tuples of vectors in  $\mathbb{R}^n$ .

#### Theorem 13.1

(a) If  $\mathbf{I}$  is the identity operator on  $\mathbb{R}^n$ , then

$$\det [\mathbf{I}] = \det (\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n) = 1.$$

- (b)  $\det$  is a linear function of each of the column vectors  $\mathbf{x}_j$ , if the others are held fixed.
- (c) If  $[A]_1$  is obtained from  $[A]$  by interchanging two columns, then
 
$$\det [A]_1 = - \det [A].$$
- (d) If  $[A]$  has two equal columns,
  - a. then  $\det [A] = 0$ .

**Proof**

- (a) If  $A = \mathbf{I}$ ,
  - then  $a(I, i) = 1$  and
  - $a(I, j) = 0$  for  $i \neq j$ .
 Hence,
 
$$\det [\mathbf{I}] = s(1, 2, 3, \dots, n) = 1$$
 Hence,
 
$$\det [\mathbf{I}] = \det (\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n) = 1.$$
 Hence part(a) of the theorem is proved.

(b) By definition,  $s(j_1, j_2, \dots, j_n) = 0$  if any two of the  $j$ 's are equal. Each of the remaining  $n!$  products in
 
$$\det [A] = s(j_1, j_2, \dots, j_n) a(1, j_1) a(2, j_2) \dots a(n, j_n),$$
 contains exactly one factor from each column. Hence  $\det$  is a linear function of each of the column vectors  $\mathbf{x}_j$ , if the others are held fixed.

(c) Let  $[A]_1$  is obtained from  $[A]$ , by interchanging two columns. Since
 
$$s(j_1, j_2, \dots, j_n) = \prod_{p < q} \text{sgn}(j_q - j_p),$$
 changes sign if any two of the  $j$ 's are interchanged,
 
$$\det [A]_1 = - \det [A].$$

(d) Let  $[A]_1$  is obtained from  $[A]$ , by interchanging two columns. Then by I,

$$\det [A]_1 = - \det [A]. \tag{1}$$

If the two columns are equal, then  $[A]_1 = [A]$ .

Therefore ,

$$\det [A]_1 = \det [A]. \tag{2}$$

From (1) and (2), we get that

$$\det [A] = 0.$$

Hence the theorem.

**Theorem 13.2**

If  $[A]$  and  $[B]$  are  $n$  by  $n$  matrices, then
 
$$\det ([B][A]) = \det [B] \det [A].$$

**Proof**

Let  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$  are the columns of  $[A]$ .

Then  $\mathbf{x}_j = \sum_{i=1}^n a(i, j) \mathbf{e}_i$ , for  $j=1, 2, \dots, n$ . ... (1)

Define  ${}_B(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) = {}_B[A] = \det ([B] [A])$ . ... (2)

The columns of  $[B][A]$  are

$$B\mathbf{x}_1, B\mathbf{x}_2, \dots, B\mathbf{x}_n.$$

Therefore,

$${}_B(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) = \det(B\mathbf{x}_1, B\mathbf{x}_2, \dots, B\mathbf{x}_n).$$

Hence  ${}_B$  has all the properties (b) to (d) of theorem13.1.

From (1),

$$\mathbf{x}_1 = \sum_{i=1}^n a(i,1)e_i .$$

Therefore using property (b) of theorem13.1, we get

$$\begin{aligned} {}_B[A] &= {}_B\left(\sum_{i=1}^n a(i,1)e_i, \mathbf{x}_2, \dots, \mathbf{x}_n\right) \\ &= \sum_{i=1}^n a(i,1) {}_B(\mathbf{e}_i, \mathbf{x}_2, \dots, \mathbf{x}_n). \end{aligned}$$

Repeating this for the column vectors  $\mathbf{x}_2, \dots, \mathbf{x}_n$ , we get

$${}_B[A] = a(i_1,1)a(i_2, 2)\dots a(i_n, n) {}_B(e_{i_1}, e_{i_2}, \dots, e_{i_n}), \quad \dots(3)$$

The sum being extended over all ordered n-tuples  $(i_1, i_2, \dots, i_n)$  with  $1 \leq i_r \leq n$ .

By (c) and (d) of theorem13.1, we get

$${}_B(e_{i_1}, e_{i_2}, \dots, e_{i_n}) = t(i_1, i_2, \dots, i_n) {}_B(\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n),$$

where  $t = 1, 0, -1$ .

Since  $[B][I] = [B]$ ,

from(2), we get that

$$\begin{aligned} {}_B(\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n) &= {}_B[I] \\ &= \det([B][I]) \\ &= \det [B]. \end{aligned}$$

Therefore ,

$${}_B(e_{i_1}, e_{i_2}, \dots, e_{i_n}) = t(i_1, i_2, \dots, i_n) \det [B].$$

Substituting this in(3), we get

$${}_B[A] = a(i_1,1)a(i_2, 2)\dots a(i_n, n) t(i_1, i_2, \dots, i_n) \det [B].$$

$$\text{i.e } \det([B][A]) = \{ a(i_1,1)a(i_2, 2)\dots a(i_n, n) t(i_1, i_2, \dots, i_n) \} \det [B], \quad \dots(4)$$

for all n by n matrices  $[A]$  and  $[B]$ .

Put  $B=I$  in (4).

Then,

$$\det([I][A]) = \{ a(i_1,1)a(i_2, 2)\dots a(i_n, n) t(i_1, i_2, \dots, i_n) \} \det [I].$$

Since  $\det [I] = 1$  and

$$[I][A] = [A], \text{ we get,}$$

$$\det [A] = a(i_1,1)a(i_2, 2)\dots a(i_n, n) t(i_1, i_2, \dots, i_n).$$

Substituting this in (4), we get,

$$\det([B][A]) = \det [A] \det [B].$$

Hence the theorem.

### Theorem13.3

A linear operator  $A$  on  $R^n$  is invertible if and only if  $\det [A] \neq 0$ .

#### Proof

Let the linear operator  $A$  on  $R^n$  be invertible.

Then  $[A][A^{-1}] = I$ .

Therefore,

$$\det ([A][A^{-1}]) = \det [I] = 1. \quad \dots(1)$$

By the theorem 13.2,

$$\det ([A][A^{-1}]) = \det [A] \det[A^{-1}]. \quad \dots(2)$$

From (1) and (2),

$$\det [A] \det[A^{-1}] = 1.$$

Hence  $\det[A] \neq 0$ .

Conversely assume that  $\det [A] \neq 0$ .

To prove that A is invertible.

Suppose if A is not invertible.

Let the column vectors of A be  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ .

Since A is not invertible,

the column vectors are dependent.

Hence there is one  $\mathbf{x}_k$  such that

$$\mathbf{x}_k + \sum_{j \neq k} c_j \mathbf{x}_j = \mathbf{0}, \text{ for certain scalars } c_j.$$

By the properties (b) and (d) of theorem 13.1,

$\mathbf{x}_k$  can be replaced by  $\mathbf{x}_k + c_j \mathbf{x}_j$  without altering the determinant if  $j \neq k$ .

Repeating this, we can replace  $\mathbf{x}_k$  by  $\mathbf{x}_k + \sum_{j \neq k} c_j \mathbf{x}_j$ , i.e by 0,

without altering the determinant.

But determinant of a matrix which has  $\mathbf{0}$  for one column is 0.

i.e  $\det [A] = 0$ , which is a contradiction.

Hence A is invertible.

Hence the theorem.

### Check your progress

Show that the determinant of the matrix of a linear operator does not depend on the basis which is used to construct the matrix.

### Answer

Suppose  $\{ \mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n \}$  and  $\{ \mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n \}$  are bases in  $\mathbb{R}^n$ .

Every linear operator A on  $\mathbb{R}^n$  determines matrices [A] and  $[A]_U$ , with entries  $a_{ij}$  and  $u_{ij}$ .

The jth column of [A] is given by

$$A\mathbf{e}_j = \sum_i a_{ij} \mathbf{e}_i, \quad j=1, 2, \dots, n \quad \dots(1)$$

and the jth column of  $[A]_U$  is given by

$$A\mathbf{u}_j = \sum_k \alpha_{kj} \mathbf{u}_k, \quad j=1, 2, \dots, n \quad \dots(2)$$

$$\text{If } \mathbf{u}_k = B\mathbf{e}_k = \sum_i b_{ik} \mathbf{e}_i, \quad \dots(3)$$

then from (2),

$$\begin{aligned} A\mathbf{u}_j &= \sum_k \alpha_{kj} B\mathbf{e}_k \\ &= \sum_k \alpha_{kj} \sum_i b_{ik} \mathbf{e}_i \\ &= \sum_i \left( \sum_k b_{ik} \alpha_{kj} \right) \mathbf{e}_i \end{aligned} \quad \dots(4)$$

From (3),

$$B\mathbf{e}_j = b_{kj}\mathbf{e}_k$$

Therefore,

$$\begin{aligned} A\mathbf{u}_j &= A B \mathbf{e}_j \\ &= A \sum_k b_{kj} \mathbf{e}_k \\ &= \sum_k b_{kj} A \mathbf{e}_k \\ &= \sum_k b_{kj} \left( \sum_i a_{ik} \mathbf{e}_i \right) \\ &= \sum_i \left( \sum_k a_{ik} b_{kj} \right) \mathbf{e}_i. \end{aligned} \quad \dots(5)$$

From (4) and (5),

$$\sum_k b_{ik} \alpha_{kj} = \sum_k a_{ik} b_{kj}$$

Hence  $[B][A]_U = [A][B]$ .

Therefore,

$$\det([B][A]_U) = \det([A][B])$$

Therefore,

$$\det[B]\det[A]_U = \det[A]\det[B] \quad \dots(6)$$

Since B is invertible,

$$\det[B] \neq 0.$$

Hence (6) implies,

$$\det[A]_U = \det[A].$$

Hence the determinant of the matrix of a linear operator does not depend on the basis which is used to construct the matrix.

### 13.3 Jacobians

If  $\mathbf{f}$  maps an open set  $E \subset \mathbb{R}^n$  into  $\mathbb{R}^n$ , and

if  $\mathbf{f}$  is differentiable at a point  $\mathbf{x} \in E$ ,

the determinant of the linear operator  $\mathbf{f}'(\mathbf{x})$  is called the Jacobian of  $\mathbf{f}$  at  $\mathbf{x}$ .

In Symbols,

$$\mathbf{J}_f(\mathbf{x}) = \det \mathbf{f}'(\mathbf{x}).$$

If  $(y_1, y_2, \dots, y_n) = \mathbf{f}(x_1, x_2, \dots, x_n)$ ,

then  $\mathbf{J}_f(\mathbf{x})$  is denoted by  $\frac{\partial(y_1, y_2, \dots, y_n)}{\partial(x_1, x_2, \dots, x_n)}$ .

### 13.4 Let us sum up

In this lesson we have seen

- the definition of determinant of the matrix of a linear operator
- Some of the properties of the determinant
- Determinant of the product of two matrices is the product of the determinants.
- The matrix of a linear operator does not depend on the basis used to construct the matrix and
- The Jacobian of a differentiable function.

### 13.5 Lesson End Activities

1. Set  $f = (f_1, f_2, f_3)$  be the vector valued function defined for each pt:  $(x_1, x_2, x_3) \in \mathbb{R}^3$  (for which  $x_1 + x_2 + x_3 \neq -1$ ) as follows:

$$F_k(x_1, x_2, x_3) = x_k / (1 + x_1 + x_2 + x_3) \quad (k=1,2,3)$$

Show that  $J_f(x_1, x_2, x_3) = (1 + x_1 + x_2 + x_3)^{-4}$

Show that  $f$  is  $1-1$  and find  $f^{-1}$  explicitly

### 13.6 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## LESSON - 14

### DERIVATIVES OF HIGHER ORDER

#### Contents

#### 14.0 Introduction

#### 14.1 Aims and objectives

#### 14.2 Derivatives of higher order

#### 14.3 Let us sum up

#### 14.4 Lesson End Activities

#### 14.5 References

#### 14.0 Introduction

In this lesson, we are going to study about the second order partial derivatives of a function.

#### 14.1 Aims and objectives

After studying this lesson you would know

- The definition of second order partial derivatives of a function
- The mean value theorem for a function of two variables and
- The necessary conditions for  $D_{ij} f = D_{ji} f$ .

#### 14.2 Derivatives of higher order

##### Definition

Suppose  $f$  is a real function defined in an open set  $E \subset \mathbb{R}^n$ , with partial derivatives  $D_1 f, D_2 f, \dots, D_n f$ .

If the functions  $D_j f$  are themselves differentiable,

then the second –order partial derivatives of  $f$  are defined by

$$D_{ij} f = D_i D_j f \quad i, j=1, 2, \dots, n.$$

If all these functions  $D_{ij} f$  are continuous in  $E$ ,

then  $f$  is said to be of class  $C''$  in  $E$ , or that  $f \in C''(E)$ .

A mapping  $f$  of  $E$  into  $\mathbb{R}^m$  is said to be of class  $C''$ , if each component of  $f$  is of class  $C''$ .

#### Theorem 14.1 Mean value theorem for functions of two variables

Suppose  $f$  is defined in an open set  $E \subset \mathbb{R}^2$ , and

$D_1 f$  and  $D_{21} f$  exist at every point of  $E$ .

Suppose  $Q \subset E$  is a closed rectangle with sides parallel to the coordinate axes,

having  $(a, b)$  and  $(a+h, b+k)$  as opposite vertices ( $h > 0, k > 0$ ).

Put

$$(f, Q) = f(a+h, b+k) - f(a+h, b) - f(a, b+k) + f(a, b).$$

Then there is a point  $(x, y)$  in the interior of  $Q$  such that

$$(f, Q) = hk(D_{21}f)(x, y).$$

##### Proof

$$\text{Let } u(t) = f(t, b+k) - f(t, b). \quad \dots(1)$$

Then,

$$\begin{aligned} (f, Q) &= f(a+h, b+k) - f(a+h, b) - f(a, b+k) + f(a, b) \\ &= u(a+h) - u(a) \\ &= hu'(x) \quad \text{for some } x \in (a, a+h) \end{aligned} \quad \dots(2)$$

[by the mean value theorem of real functions of single variables,

“If  $f$  is a real continuous function on  $[a, b]$  which is differentiable in  $(a, b)$ , then there is a point  $x \in (a, b)$  at which

$$f(b) - f(a) = (b-a)f'(x). \quad [ ]$$

From (1),

$$u'(t) = D_1f(t, b+k) - D_1f(t, b).$$

Substituting in (2),

$$(f, Q) = h[ D_1f(x, b+k) - D_1f(x, b)]. \quad \dots(3)$$

Again applying the mean value theorem,

there exists a point  $y \in (b, b+k)$  such that

$$D_1f(x, b+k) - D_1f(x, b) = kD_{21}f(x, y)$$

Substituting in(3), we get that

$$(f, Q) = hkD_{21}f(x, y).$$

Hence there is a point  $(x, y)$  in the interior of  $Q$  such that

$$(f, Q) = hk(D_{21}f)(x, y).$$

Hence the theorem

**Theorem14.2**

Suppose  $f$  is defined in an open set  $E \subset \mathbb{R}^2$ , suppose that  $D_1f, D_{21}f$ , and  $D_{22}f$  exist at every point of  $E$ , and  $D_{21}f$  is continuous at some point  $(a, b) \in E$ .

Then  $D_{12}f$  exists at  $(a, b)$  and

$$(D_{12}f)(a, b) = (D_{21}f)(a, b).$$

**Proof**

Let  $A = (D_{21}f)(a, b)$ .

Choose  $\epsilon > 0$ .

Let  $Q \subset E$  be a closed rectangle with sides parallel to the coordinate axes, having  $(a, b)$  and  $(a+h, b+k)$  as opposite vertices ( $h > 0, k > 0$ ).

Then by theorem14.1,

$$\begin{aligned} (f, Q) &= f(a+h, b+k) - f(a+h, b) - f(a, b+k) + f(a, b) \\ &= hk(D_{21}f)(x, y), \quad \text{for some point } (x, y) \in Q. \end{aligned} \quad \dots(1)$$

If  $h$  and  $k$  are sufficiently small, then we have

$$|A - (D_{21}f)(x, y)| < \epsilon \quad \text{for all } (x, y) \in Q. \quad \dots(2).$$

From (1) and (2), we get

$$\left| A - \frac{\Delta(f, Q)}{hk} \right| < \epsilon.$$

i.e  $\left| \frac{\Delta(f, Q)}{hk} - A \right| < \epsilon$

i.e  $\left| \frac{f(a+h, b+k) - f(a+h, b) - f(a, b+k) + f(a, b)}{hk} - A \right| < \epsilon$

i.e  $\left| \frac{\frac{f(a+h, b+k) - f(a+h, b)}{k} - \frac{f(a, b+k) - f(a, b)}{k}}{h} - A \right| < \epsilon$

Keeping  $h$  fixed, let  $k > 0$ .

Since  $D_{22}f$  exists in  $E$ , the last inequality implies that

$$\left| \frac{D_2 f(a+h, b) - D_2 f(a, b)}{h} - A \right| \dots(3)$$

Since  $h$  was arbitrary, and since (3) holds for all sufficiently small  $h > 0$ , (3) implies that  $(D_{21}f)(a, b)$  exists and

$$(D_{21}f)(a, b) = A.$$

Hence  $(D_{12}f)(a, b) = (D_{21}f)(a, b)$ .

### Check your progress.

Put  $f(0, 0) = 0$ . and

$$f(x, y) = \frac{xy(x^2 - y^2)}{x^2 + y^2} \quad \text{if } (x, y) \neq (0, 0).$$

Prove that (a)  $f$ ,  $D_1f$ ,  $D_2f$  are continuous in  $\mathbb{R}^2$ .

(b)  $D_{12}f$  and  $D_{21}f$  exist at every point of  $\mathbb{R}^2$ , and are continuous except at  $(0, 0)$

(c)  $(D_{12}f)(0, 0) = 1$  and  $(D_{21}f)(0, 0) = -1$ .

### 14.3 Let us sum up

In this lesson we have seen

- The definition of second order partial derivatives which we can extend to higher order partial derivatives.
- Mean value theorem for functions of two variables. and
- Conditions to be satisfied in order that the partial derivatives  $D_{12}f$  and  $D_{21}f$  are equal.

### 14.4 Lesson End Activities

1. if  $f(x, y) = \frac{xy}{x^2 + y^2}$  if  $(x, y) \neq (0, 0) = 0$  if  $(x, y) = (0, 0)$ . Prove that both the partial derivatives exist at  $(0, 0)$  but the function is not continuous at  $(0, 0)$
2. Expand  $x^3y + 3y - z$  in powers of  $(x-1)$  and  $(y+2)$ .

### 14.5 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Wiley and Sons, New York, 1976.
2. W. Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## LESSON - 15

### DIFFERENTIATION OF INTEGRALS

#### 15.0 Introduction

##### 15.1 Aims and objectives

##### 15.2 Differentiation of Integrals

##### 15.3 Let us sum up

##### 15.4 References

#### 15.0 Introduction

In this lesson, we are going to study about differentiation of integrals. Suppose  $\varphi$  is a function of two variables which can be integrated with respect to one variable and which can be differentiated with respect to another variable. We are going to see under what conditions the result will be the same if these two limit process are carried out in the opposite order.

#### 15.1 Aims and objectives

After studying this lesson, you would know

- Under what conditions the following equation is true?.

$$\frac{d}{dt} \int_a^b \varphi(x, t) dx = \int_a^b \frac{\partial \varphi}{\partial t}(x, t) dx$$

Note:

It will be convenient to use the notation

$$f^t(x) = \varphi(x, t)$$

Thus  $f^t$  is a function of one variable, for each  $t$ .

#### 15.2 Differentiation of Integrals

##### Theorem 15.1

Suppose

- $\varphi(x, t)$  is defined for  $a \leq x \leq b, c \leq t \leq d$ ;
- $\alpha$  is an increasing function on  $[a, b]$ ;
- $f^t \in R(\alpha)$  for every  $t \in [c, d]$ ;
- $c < s < d$ , and to every  $\epsilon > 0$  corresponds a  $\delta > 0$  such that
 
$$|(D_{\alpha} \varphi)(x, t) - (D_{\alpha} \varphi)(x, s)| < \epsilon$$
 for all  $x \in [a, b]$  and for all  $t \in (s - \delta, s + \delta)$ .

Define

$$f(t) = \int_a^b \varphi(x, t) d\alpha(x) \quad c \leq t \leq d.$$

Then  $(D_{\alpha} \varphi)^s \in R(\alpha)$ ,

$f'(s)$  exists, and

$$f'(s) = \int_a^b (D_{\alpha} \varphi)(x, s) d\alpha(x).$$

##### Proof

For  $0 < |t - s| < \delta$ ,

consider the difference quotients

$$f^t(x) = \frac{\varphi(x, t) - \varphi(x, s)}{t - s}.$$

Since  $f^t \in R(\alpha)$  for every  $t \in [c, d]$ ,

$t \in R(\cdot)$ , for each  $t \in [c, d]$ .

By mean value theorem,

there exists a  $u$  between  $s$  and  $t$  such that

$$f(x, t) - f(x, s) = D_2 f(x, u)(t - s)$$

Hence to each  $(x, t)$ ,

there exists a  $u$  between  $s$  and  $t$  such that

$$f(x, t) = D_2 f(x, u) \quad \text{whenever } 0 < |t - s| < \delta. \quad \dots(1)$$

Hence assumption (d) implies that

$$|f(x, t) - (D_2 f)(x, s)| < \epsilon \quad \dots(2)$$

whenever  $a \leq x \leq b$  and  $0 < |t - s| < \delta$ .

Consider

$$\begin{aligned} \frac{f(t) - f(s)}{t - s} &= \frac{1}{t - s} \left[ \int_a^b \varphi(x, t) d\alpha(x) - \int_a^b \varphi(x, s) d\alpha(x) \right] \\ &= \int_a^b \frac{\varphi(x, t) - \varphi(x, s)}{t - s} d\alpha(x) \\ &= \int_a^b \psi(x, t) d\alpha(x) \quad \dots(3) \end{aligned}$$

From(2), we get that

$$|\psi(x, t) - (D_2 \varphi)(x, s)| < \epsilon, \text{ uniformly on } [a, b], \text{ as } |t - s| < \delta. \quad \dots(4)$$

Also each  $t \in R(\cdot)$ .

$$\begin{aligned} \text{Hence } f'(s) &= \lim_{t \rightarrow s} \frac{f(t) - f(s)}{t - s} \\ &= \lim_{t \rightarrow s} \int_a^b \psi(x, t) d\alpha(x) \\ &= \int_a^b \lim_{t \rightarrow s} \psi(x, t) d\alpha(x) \\ &= \int_a^b (D_2 \varphi)(x, s) d\alpha(x) \quad \text{[from (4) and theorem 7.1]} \end{aligned}$$

Hence the theorem.

**Example**

Define  $f(t) = \int_{-\infty}^{\infty} e^{-x^2} \cos(xt) dx$

And  $g(t) = - \int_{-\infty}^{\infty} x e^{-x^2} \sin(xt) dx$ , for  $-\infty < t < \infty$ .

Both integrals exist since the absolute values of the integrands are at most  $\exp(-x^2)$  and  $|x| \exp(-x^2)$ , respectively.

Also  $g$  can be obtained from  $f$  by differentiating the integrand of  $f(t)$  with respect to  $t$ .

Claim:  $f'(t) = g(t)$   $-\infty < t < \infty$ .

Proof

By integrating the right hand side, we can prove that

$$\frac{\cos(\alpha + \beta) - \cos \alpha}{\beta} + \sin \alpha = \frac{1}{\beta} \int_{\alpha}^{\alpha + \beta} (\sin \alpha - \sin t) dt, \text{ for } \beta > 0.$$

Therefore

$$\begin{aligned} \left| \frac{\cos(\alpha + \beta) - \cos \alpha}{\beta} + \sin \alpha \right| &= \left| \frac{1}{\beta} \int_{\alpha}^{\alpha + \beta} (\sin \alpha - \sin t) dt \right| \\ &= \frac{1}{\beta} \int_{\alpha}^{\alpha + \beta} |\sin \alpha - \sin t| dt \end{aligned}$$

$$\frac{1}{\beta} \int_{\alpha}^{\alpha+\beta} |\alpha - t| dt$$

$$\frac{1}{\beta} \int_{\alpha}^{\alpha+\beta} (t - \alpha) dt$$

$$\frac{1}{\beta} \left[ \frac{(t - \alpha)^2}{2} \right]_{\alpha}^{\alpha+\beta}$$

$$/2$$

Similar result can be obtained if  $h < 0$ .

Therefore

$$\left| \frac{\cos(\alpha + \beta) - \cos \alpha}{\beta} + \sin \alpha \right| \leq |h| \quad \text{for all } \alpha.$$

Fix  $t$ , and fix  $h > 0$ .

Substituting  $t = xt$ ,  $t = xh$  in the above inequality, we get

$$\left| \frac{\cos(t + h)x - \cos tx}{xh} + \sin tx \right| \leq |x| |h|$$

Therefore,

$$\left| \frac{\cos(t + h)x - \cos tx}{h} + x \sin tx \right| \leq |x|^2 |h|. \quad \dots(1)$$

Consider

$$\left| \frac{f(t + h) - f(t)}{h} - g(t) \right|$$

$$= \left| \frac{\int_{-\infty}^{\infty} e^{-x^2} \cos(t + h)x dx - \int_{-\infty}^{\infty} e^{-x^2} \cos(tx) dx}{h} + \int_{-\infty}^{\infty} x e^{-x^2} \sin(tx) dx \right|$$

$$= \left| \int_{-\infty}^{\infty} e^{-x^2} \left( \frac{\cos(t + h)x - \cos(tx)}{h} + x \sin(tx) \right) dx \right|$$

$$\int_{-\infty}^{\infty} e^{-x^2} \left| \left( \frac{\cos(t + h)x - \cos(tx)}{h} + x \sin(tx) \right) \right| dx$$

$$\int_{-\infty}^{\infty} e^{-x^2} |x|^2 |h| dx$$

$$|h| \int_{-\infty}^{\infty} e^{-x^2} x^2 dx > 0 \quad \text{as } h > 0.$$

Hence  $f'(t) = g(t) \quad - < t < .$

Hence the required result.

### 15.3 Let us sum up

In this lesson, we have seen

- under what conditions a function of two variables can be integrated with respect to one variable and can be differentiated with respect to another variable and
- an example .

## 15.4 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## Unit 4 Lebesgue Measure and Lebesgue integral.

### Lesson 16 Outer measure, Measurable sets and Lebesgue Measure

#### 16.0 Introduction

#### 16.1 Aims and Objectives

#### 16.2 Outer measure

#### 16.3 Measurable sets

#### 16.4 Lebesgue measure

#### 16.5 Let us sum up

#### 16.6 Lesson End Activities

#### 16.7 References

#### 16.0 Introduction

In this lesson, we are going to study about Lebesgue outer measure of a set, measurable sets and Lebesgue measure, their important properties and littlewood's First principle.

#### 16.1 Aims and Objectives

After studying this lesson, you would know

- The definition of outer measure of sets
- Outer measure of an interval
- Some important properties of Outer measure
- The definition of Measurable sets
- Measure of countable union of measurable sets
- Measure of Countable intersection of measurable sets and
- Littlewood's First Principle.

#### 16.2 Outer measure

##### Definition

A collection  $\mathcal{M}$  of sets is called a  $\sigma$ -algebra or a Borel set if

- (i) Every union of a countable collection of sets in  $\mathcal{M}$  is again in  $\mathcal{M}$  and
- (ii) If  $A \in \mathcal{M}$ , then  $\tilde{A} \in \mathcal{M}$ .

##### Definition : Outer measure

Let  $A$  be set of real numbers.

The outer measure  $m^*(A)$  of  $A$  is defined as

$$m^*(A) = \inf_{A \subset \cup I_n} \sum l(I_n)$$

where  $l(I_n)$  denotes the length of  $I_n$ .

##### Theorem 16.1

- (a) If  $A \subset (a, b)$ , then  $m^*(A) = (b - a)$ .
- (b) If  $A \subset B$ , then  $m^*(A) \leq m^*(B)$
- (c)  $m^*({x}) = 0$ , where  $x$  is a real number.
- (d)  $m^*(\phi) = 0$

##### Proof

- (a) Let  $U = \{\cup I_n / A \subset \cup I_n\}$

Since  $A \subset (a, b)$ ,  
 $(a, b) \in U$ .

Therefore,

$$\inf_{A \subset \cup I_n} \sum l(I_n) = l(a, b) = (b - a)$$

Therefore,

$$m^*(A) = (b - a).$$

(b) Let  $A \subset B$ .

Let  $U = \{ \cup I_n / A \subset \cup I_n \}$  and

$$V = \{ \cup J_n / B \subset \cup J_n \}$$

Since  $A \subset B$ ,

$$A \subset \cup J_n.$$

Therefore every element of  $V$  is an element of  $U$ .

Hence  $V \subset U$ .

Hence,

$$\inf_{A \subset \cup I_n} \sum l(I_n) \leq \inf_{B \subset \cup J_n} \sum l(J_n)$$

Hence  $m^*(A) \leq m^*(B)$ .

(c) Let  $x \in \mathbb{R}$ .

For any  $\epsilon > 0$ ,

$$\{x\} \subset (x - \epsilon, x + \epsilon) \text{ and}$$

$$l(x - \epsilon, x + \epsilon) = 2\epsilon.$$

Therefore by (a),

$$m^*({x}) \leq 2\epsilon.$$

Since  $\epsilon$  was arbitrary,

$$\text{this implies } m^*({x}) = 0.$$

(e) Let  $x \in \mathbb{R}$ .

Then  $\emptyset \subset \{x\}$

Therefore by (b),

$$m^*(\emptyset) \leq m^*({x})$$

Since  $m^*({x}) = 0$ ,

$$m^*(\emptyset) = 0.$$

Hence the theorem.

### Theorem 16.2

The outer measure of an interval is its length.

#### Proof

**Case 1.** Let  $I$  be a finite closed interval, say  $[a, b]$ .

Therefore,

$$I \subset (a - \epsilon, a + \epsilon) \text{ for any } \epsilon > 0.$$

Therefore,

$$m^*(I) \leq b - a + 2\epsilon \quad [\text{by the theorem 16.1(a)}]$$

Since this is true for every  $\epsilon > 0$ ,

$$m^*(I) = b - a \quad \dots (1)$$

Let  $I \subset \bigcup_{n=1}^{\infty} I_n$ .

By Heine – Borel theorem,

There exists a finite subcollection from  $\{ I_n \}$ , say  $I_1, I_2, \dots, I_m$ , which covers  $I$ , and

$$\sum_{k=1}^m l(I_k) = \sum_{n=1}^{\infty} l(I_n).$$

Since  $I=[a, b] \subset I_1 \cup I_2 \cup \dots \cup I_m$ ,  
 $a \in I_l$  for some  $l$  with  $1 \leq l \leq m$ .

Let  $I_l$  be  $(a_1, b_1)$ .

Therefore  $a_1 < a < b_1$ .

If  $b < b_1$ ,

then  $a_1 < a < b < b_1$ . i.e  $(a, b) \subset I_l$ .

Therefore  $l(I_l) = b - a$ .

$$\text{Therefore } \sum_{n=1}^{\infty} l(I_n) = l(I_l) = b - a.$$

If  $b_1 < b$ ,

then  $b_1 \in [a, b]$  but  $b_1 \notin I_l$ .

Therefore  $b_1 \in I_t$  for some  $t$ ,  $1 \leq t \leq m$ .

Let  $I_t$  be  $(a_2, b_2)$ .

Therefore  $a_2 < b_1 < b_2$ .

If  $b < b_2$ ,

then  $a_1 < a < b_1 < b < b_2$ .

Therefore,

$$\begin{aligned} \sum_{n=1}^{\infty} l(I_n) &= l(I_l) + l(I_t) \\ &= (b_2 - a_2) + (b_1 - a_1) \\ &= (b_2 - a_1) + (b_1 - a_2) \\ &> (b_2 - a_1) \\ &= b - a. \end{aligned}$$

If  $b_2 < b$ ,

continuing in the same way, we obtain a sequence

$(a_1, b_1), (a_2, b_2), \dots, (a_k, b_k)$  from  $\{I_n\}_{n=1}^m$  such that  $a_i < b_{i-1} < b_i$ .

Since  $\{I_n\}_{n=1}^m$  is a finite collection, this process must terminate with some interval say

$(a_r, b_r)$ .

But it terminates only if  $b \in (a_r, b_r)$ .

Therefore,

$$\begin{aligned} \sum_{n=1}^{\infty} l(I_n) &= \sum_{i=1}^r l(a_i, b_i) \\ &= (b_r - a_r) + (b_{r-1} - a_{r-1}) + \dots + (b_1 - a_1) \\ &= b_r - (a_r - b_{r-1}) - \dots - (a_2 - b_1) - a_1 \\ &= b_r - a_1. \quad [\text{since } a_i < b_{i-1}, 1 \leq i \leq r. \\ &= b - a. \end{aligned}$$

Therefore,  $\inf \sum_{n=1}^{\infty} l(I_n) = b - a$ .

Hence  $m^*(I) = b - a \dots(2)$

From (1) and (2),

$$m^*(I) = b - a.$$

Case 2: Let  $I$  be any finite interval

i.e  $I = [a, b]$ , or  $(a, b)$ , or  $(a, b]$ .

Then there exists an interval  $J = [a + \epsilon/4, b - \epsilon/4]$  such that  $J \subset I$ .

$$\begin{aligned} \text{But } m^*(J) &= l(J) \\ &= (b - \epsilon/4) - (a + \epsilon/4) \quad [\text{by case 1}] \\ &= b - a - \epsilon/2 \\ &= l(I) - \epsilon/2 \\ &> l(I) - \epsilon \end{aligned}$$

Then for each  $\epsilon > 0$ ,

$$\begin{aligned} l(I) - \epsilon &< l(J) \\ &= m^*(J) \\ &= m^*(I) \\ &= m^*(\bar{I}) \\ &= l(\bar{I}) \\ &= l(I). \end{aligned}$$

Therefore

$$l(I) - \epsilon < m^*(I) \leq l(I).$$

Since this is true for every  $\epsilon > 0$ , we get that

$$m^*(I) = l(I).$$

case 3. Let  $I$  be any infinite interval.

Then given any real number  $\epsilon$ ,

there is a closed interval  $J \subset I$  with  $l(J) = \epsilon$ .

Hence,

$$m^*(I) \geq m^*(J) = l(J) = \epsilon.$$

Therefore,

$$m^*(I) \geq \epsilon \quad \text{for each } \epsilon > 0.$$

This implies,

$$m^*(I) = \infty = l(I).$$

Hence in all cases  $m^*(I) = l(I)$ .

Hence the theorem.

### Theorem 16.3

Let  $\{A_n\}_{n=1}^{\infty}$  be a countable collection of sets of real numbers.

$$\text{Then } m^*\left(\bigcup_{n=1}^{\infty} A_n\right) \leq \sum_{n=1}^{\infty} m^*(A_n).$$

#### Proof

Suppose  $m^*(A_n) = \infty$  for some  $n$ .

$$\text{Then } \sum_{n=1}^{\infty} m^*(A_n) = \infty.$$

$$\text{Hence } m^*\left(\bigcup_{n=1}^{\infty} A_n\right) \leq \sum_{n=1}^{\infty} m^*(A_n).$$

Hence the theorem is true.

So assume that  $m^*(A_n) < \epsilon$ , for all  $n$ .

i.e  $\inf_{A_n \subset \cup I_{n,i}} \sum l(I_{n,i}) < \epsilon$ , for all  $n$ .

Hence for given  $\epsilon > 0$ ,

there exist countable collection  $\{ I_{n,i} \}_i$  of open intervals such that

$$A_n \subset \bigcup_i I_{n,i} \text{ and}$$

$$\sum_i l(I_{n,i}) < m^*(A_n) + 2^{-n} \epsilon \quad \dots(1)$$

Since a union of a countable number of countable collections is countable,

$\{ I_{n,i} \}_{n,i}$  is countable and

$$\bigcup_{n=1}^{\infty} A_n \subset \bigcup_{n,i} I_{n,i}$$

Hence

$$m^*\left(\bigcup_i I_{n,i}\right) \leq \sum_{n,i} l(I_{n,i})$$

$$= \sum_n \sum_i l(I_{n,i})$$

$$< \sum_n (m^*(A_n) + 2^{-n} \epsilon)$$

$$< \sum_n m^*(A_n) + \sum_n (2^{-n} \epsilon)$$

$$= \sum_n m^*(A_n) + \epsilon$$

Since  $\epsilon$  was an arbitrary positive number,

$$m^*\left(\bigcup_{n=1}^{\infty} A_n\right) \leq \sum_{n=1}^{\infty} m^*(A_n).$$

Hence the theorem.

Remark:

- This property is called countable subadditivity of  $m^*$ .
- Clearly  $m^*$  satisfies the finite subadditivity

i.e  $m^*\left(\bigcup_{m=1}^n A_m\right) \leq \sum_{m=1}^n m^*(A_m).$

**Theorem16.4**

If  $A$  is countable, prove that  $m^*(A) = 0$ .

**Proof**

Let  $A = \{ x_1, x_2, \dots, x_n, \dots \}$ .

$$m^*(A) = m^*\left[\bigcup_i \{x_i\}\right]$$

$$= \sum_i m^*({x_i}) \quad \text{[by theorem16.3]}$$

$$= 0 \quad \text{[since } m^*({x}) = 0, \text{ for all real } x]$$

Hence  $m^*(A) = 0$ .

Hence the theorem.

**Theorem16.5**

The set  $[0, 1]$  is not countable.

**Proof**

On the contrary ,

assume that  $[0, 1]$  is countable.

Then by theorem 16.4,

$$m^*([0, 1]) = 0$$

But by theorem 16.2,

$$m^*([0, 1]) = (1 - 0) = 1$$

which is a contradiction.

Hence  $[0, 1]$  is not countable.

**Check your progress**

1. If  $Q$  denotes the set of all rational numbers, prove that  $m^*(Q) = 0$

Given any set  $A$  and  $\epsilon > 0$ , Prove that

there is an open subset  $U$  such that  $A \subset U$  and

$$m^*(U) = m^*(A) + \epsilon.$$

2.  $m^*$  is translation invariant i.e  $m^*(A+x) = m^*(A)$ , for all  $x \in \mathbb{R}$

3. Prove that if  $m^*(A) = 0$ , then  $m^*(A \cup B) = m^*(B)$ .

4. Let  $A$  be the set of all rational numbers between 0 and 1.

Let  $\{I_n\}$  be a finite collection of open intervals covering  $A$ .

Then  $\sum l(I_n) > 1$ .

**16.3 Measurable sets****Definition**

A subset  $E$  of  $\mathbb{R}$  is said to be measurable,

if for each subset  $A$  of  $\mathbb{R}$ , we have

$$m^*A = m^*(A \cap E) + m^*(A \cap \tilde{E})$$

where  $\tilde{E} = \mathbb{R} - E$ . i.e  $\tilde{E}$  denotes the complement of  $E$  in  $\mathbb{R}$ .

Remark:

1. Since we are using Lebesgue outer measure  $m^*$ , the measurable sets are called Lebesgue measurable sets.

2. Since  $A = (A \cap E) \cup (A \cap \tilde{E})$ ,

$$m^*A = m^*(A \cap E) + m^*(A \cap \tilde{E})$$

Therefore to show that  $E$  is measurable,

it is enough to show that

$$m^*A = m^*(A \cap E) + m^*(A \cap \tilde{E}).$$

3. Since the definition of measurability is symmetric in  $E$  and  $\tilde{E}$ ,  $\tilde{E}$  is measurable whenever  $E$  is measurable.

**Theorem 16.6**

$E$  and  $R$  are measurable sets.

**Proof**

If  $E = \mathbb{R}$ ,

then  $A \cap \mathbb{R} = A$  and

$$A \cap \tilde{\mathbb{R}} = A \cap \emptyset$$

$$= \emptyset.$$

Therefore,

$$m^*(A \cap \mathbb{R}) = m^*A \text{ and}$$

$$m^*(A \cap \tilde{R}) = m^* \quad = 0.$$

Therefore,

$$\begin{aligned} m^*A &= m^*(A \cap R) \\ &= m^*(A \cap R) + m^*(A \cap \tilde{R}) \end{aligned}$$

Hence R is measurable.

Therefore ,

$\tilde{R}$  is also measurable.  
i.e  $\tilde{R}$  is measurable.

Hence R and  $\tilde{R}$  are measurable sets.

Hence the theorem.

**Theoem16.7**

If  $m^*E = 0$ , then E is measurable.

**Proof**

Let A be any subset of R.

Then  $A \cap E \subset E$ .

Therefore,

$$m^*(A \cap E) = m^*E = 0.$$

Therefore,

$$m^*(A \cap E) = 0. \quad \dots(1)$$

But  $A \cap \tilde{E} \subset A$ .

Therefore,

$$m^*A = m^*(A \cap \tilde{E}) \quad \dots(2)$$

From (1) and (2), we get

$$m^*A = m^*(A \cap E) + m^*(A \cap \tilde{E}).$$

Hence E is measurable.

Hence the theorem.

**Theorem16.8**

If  $E_1$  and  $E_2$  are measurable sets , then  $E_1 \cup E_2$  is also measurable.

**Proof**

Let A be any subset of R.

Since  $E_2$  is measurable,

$$m^*A = m^*(A \cap E_2) + m^*(A \cap \tilde{E}_2).$$

Replacing A by  $A \cap \tilde{E}_1$ ,

$$m^*(A \cap \tilde{E}_1) = m^*(A \cap \tilde{E}_1 \cap E_2) + m^*(A \cap \tilde{E}_1 \cap \tilde{E}_2). \quad \dots(1)$$

Consider,

$$\begin{aligned} A \cap (E_1 \cup E_2) &= (A \cap E_1) \cup (A \cap E_2) \\ &= (A \cap E_1) \cup (A \cap E_2 \cap R) \\ &= (A \cap E_1) \cup (A \cap E_2 \cap (\tilde{E}_1 \cup E_1)) \\ &= (A \cap E_1) \cup [(A \cap E_2 \cap \tilde{E}_1) \cup (A \cap E_2 \cap E_1)] \\ &= (A \cap E_1) \cup (A \cap E_2 \cap \tilde{E}_1) \\ &\quad [\text{since } A \cap E_2 \cap E_1 \subset A \cap E_1] \end{aligned}$$

Therefore,

$$m^*(A \cap (E_1 \cup E_2)) = m^*(A \cap E_1) + m^*(A \cap E_2 \cap \tilde{E}_1)$$

Adding  $m^*(A \cap \tilde{E}_1 \cap \tilde{E}_2)$  to both sides,

$$\begin{aligned} m^*(A \cap (E_1 \cup E_2)) + m^*(A \cap \tilde{E}_1 \cap \tilde{E}_2) &= m^*(A \cap E_1) + m^*(A \cap E_2 \cap \tilde{E}_1) + m^*(A \cap \tilde{E}_1 \cap \tilde{E}_2) \\ &= m^*(A \cap E_1) + m^*(A \cap \tilde{E}_1) \quad [\text{using (1)}] \\ &= m^*A \quad [\text{Since } E_1 \text{ is measurable.}] \end{aligned}$$

Therefore ,

$$\begin{aligned} m^*A &= m^*(A \cap (E_1 \cup E_2)) + m^*(A \cap \tilde{E}_1 \cap \tilde{E}_2) \\ m^*A &= m^*(A \cap (E_1 \cup E_2)) + m^*(A \cap (\tilde{E}_1 \cup \tilde{E}_2)) \end{aligned}$$

Hence  $E_1 \cup E_2$  is measurable.

Hence the theorem.

**Check your progress**

1. Prove that the family  $M$  of measurable sets is an algebra.
2. If  $E_1, E_2, \dots, E_n$  are measurable, prove that  $E_1 \cup E_2 \cup \dots \cup E_n$  is measurable.
3. If  $E_1$  and  $E_2$  are measurable sets , then prove that  $E_1 \cup E_2$  is also measurable.

**Theorem16.9**

Let  $A$  be any set and  $E_1, E_2, \dots, E_n$  a finite sequence of disjoint measurable sets,

$$\text{Then } m^*(A \cap \left[ \bigcup_{i=1}^n E_i \right]) = \sum_{i=1}^n m^*(A \cap E_i).$$

**Proof**

We prove this theorem , by induction on  $n$ .

When  $n=1$ , the result is obvious.

Assume that the result is true for  $n-1$  sets ( $n \geq 2$ ).

$$\text{i.e } m^*(A \cap \left[ \bigcup_{i=1}^{n-1} E_i \right]) = \sum_{i=1}^{n-1} m^*(A \cap E_i). \quad \dots(1)$$

Let  $E_1, E_2, \dots, E_n$  be  $n$  disjoint measurable sets.

$$\text{Then } A \cap \left[ \bigcup_{i=1}^n E_i \right] \cap E_n = A \cap E_n \quad \dots(2)$$

$$[\text{since } E_n \subset \left[ \bigcup_{i=1}^n E_i \right]]$$

$$A \cap \left[ \bigcup_{i=1}^n E_i \right] \cap \tilde{E}_n = A \cap [(E_1 \cap \tilde{E}_n) \cup \dots \cup (E_n \cap \tilde{E}_n)]$$

Since  $E_1, E_2, \dots, E_n$  are disjoint ,

$E_i \subset E_n$  for  $i < n$ , and

$E_n \cap \tilde{E}_n = \emptyset$ , we get

$$\begin{aligned} A \cap \left[ \bigcup_{i=1}^n E_i \right] \cap \tilde{E}_n &= A \cap [E_1 \cup E_2 \cup \dots \cup E_{n-1} \cup \emptyset] \\ &= A \cap \left[ \bigcup_{i=1}^{n-1} E_i \right] \quad \dots(3) \end{aligned}$$

Since  $E_n$  is measurable,

$$m^*A = m^*(A \cap E_n) + m^*(A \cap \tilde{E}_n).$$

Replacing A by  $A \cap \left[ \bigcup_{i=1}^n E_i \right]$ , we get,

$$\begin{aligned} m^*(A \cap \left[ \bigcup_{i=1}^n E_i \right]) &= m^*(A \cap \left[ \bigcup_{i=1}^n E_i \right] \cap E_n) + m^*(A \cap \left[ \bigcup_{i=1}^n E_i \right] \cap \tilde{E}_n). \\ &= m^*(A \cap E_n) + m^*(A \cap \left[ \bigcup_{i=1}^{n-1} E_i \right]) \quad \text{[from (2) and (3)]} \\ &= m^*(A \cap E_n) + \sum_{i=1}^{n-1} m^*(A \cap E_i) \quad \text{[from(1)]} \end{aligned}$$

Hence  $m^*(A \cap \left[ \bigcup_{i=1}^n E_i \right]) = \sum_{i=1}^n m^*(A \cap E_i)$ .

Hence by induction the theorem follows.

**Remark:**

$m^*$  is finitely additive.

**Proof:**

In the above theorem, take  $A = R$ .

Then,

$$m^*(R \cap \left[ \bigcup_{i=1}^n E_i \right]) = \sum_{i=1}^n m^*(R \cap E_i).$$

Hence,

$$m^*\left(\left[ \bigcup_{i=1}^n E_i \right]\right) = \sum_{i=1}^n m^* E_i.$$

Hence the result.

**Defintion**

A collection T of subsets of a set X is called a  $\sigma$ - algebra if

- (i)  $A \in T$  implies  $\tilde{A} \in T$ .
- (ii) If  $A = \bigcup_{n=1}^{\infty} A_n$  and if  $A_n \in T$ , for every n, then  $A \in T$ .

**Theorem16.10**

The collection M of measurable sets is a  $\sigma$ - algebra.

**Proof**

If E is a measurable set,

then we know that  $\tilde{E}$  is also measurable.

Hence if  $E \in M$ , then  $\tilde{E} \in M$ .

Hence the first condition of the definition of  $\sigma$ - algebra is satisfied.

To prove the second condition:

Let  $E_n \in M, n=1,2,3,\dots$

Let  $E = \bigcup_{n=1}^{\infty} E_n$ .

Claim: There is a set  $F_n \in \mathcal{M}$ ,  $n=1,2,3,\dots$  such that  $F_i \cap F_j = \phi$  for  $i \neq j$  and

$$\bigcup_{i=1}^{\infty} E_i = \bigcup_{i=1}^{\infty} F_i .$$

To prove the claim,

Take  $F_1 = E_1$ .

For  $n > 1$ ,

$$\begin{aligned} \text{define } F_n &= E_n - \{ E_1 \cup E_2 \cup \dots \cup E_{n-1} \} \\ &= E_n \cap \tilde{E}_1 \cap \dots \cap \tilde{E}_{n-1}. \end{aligned}$$

Since  $\mathcal{M}$  is an algebra,

$$E_i \in \mathcal{M} \text{ implies } \tilde{E}_i \in \mathcal{M} \quad 1 \leq i \leq n-1.$$

This implies,

$$E_n \cap \tilde{E}_1 \cap \dots \cap \tilde{E}_{n-1} \in \mathcal{M}.$$

This implies,

$$F_n \in \mathcal{M}, \quad n=1,2,3,\dots$$

Also  $F_n \subset E_n$ , for all  $n$ .

Let  $m < n$ .

Suppose  $m < n$ .

Then ,

$$\begin{aligned} F_m \cap F_n &\subset E_m \cap F_n = E_m \cap (E_n \cap \tilde{E}_1 \cap \dots \cap \tilde{E}_{n-1}) \\ &= E_n \cap \tilde{E}_1 \cap \dots \cap \tilde{E}_{n-1} \cap E_m \\ &= \phi . \end{aligned}$$

Hence if  $m \neq n$ ,

$$\text{then } F_m \cap F_n = \phi .$$

Since  $F_i \subset E_i$ , for all  $i$ ,

$$\bigcup_{i=1}^{\infty} F_i \subset \bigcup_{i=1}^{\infty} E_i . \quad \dots(1)$$

Now, let  $x \in \bigcup_{i=1}^{\infty} E_i$ .

This implies,

$$x \in E_n \text{ for some } n.$$

Let  $m$  be the smallest value of  $n$  such that  $x \in E_m$ .

Then  $x \notin E_i$ , for  $i < m$ .

This implies that,

$$x \in \tilde{E}_i, \text{ for } i < m.$$

Therefore,

$$x \in E_m \cap \tilde{E}_{m-1} \cap \dots \cap \tilde{E}_1 .$$

Therefore,

$$x \in F_m \subset \bigcup_{i=1}^{\infty} F_i .$$

Hence  $\bigcup_{i=1}^{\infty} E_i \subset \bigcup_{i=1}^{\infty} F_i \quad \dots(2)$

From(1) and (2), we get that

$$\bigcup_{i=1}^{\infty} E_i = \bigcup_{i=1}^{\infty} F_i .$$

Hence the claim is proved.

i.e if E is the union of a countable collection of measurable sets, it must be the union of pairwise disjoint measurable sets.

$$\text{Let } A_n = \bigcup_{i=1}^n E_i .$$

Then  $A_n \in \mathcal{M}$ , [Since  $\mathcal{M}$  is an algebra and  $E_i \in \mathcal{M}$  ]

Hence  $A_n$  is measurable.

Also  $A_n \subset E$ .

Hence  $\tilde{A}_n \supset \tilde{E}$

Therefore,

$A \cap \tilde{A}_n \supset A \cap \tilde{E}$ , for any set A in R.

$$m^*(A \cap \tilde{A}_n) \geq m^*(A \cap \tilde{E}) \quad \dots(3)$$

Since  $A_n$  is measurable,

$$\begin{aligned} m^*A &= m^*(A \cap A_n) + m^*(A \cap \tilde{A}_n) \\ &= m^*(A \cap A_n) + m^*(A \cap \tilde{E}) \\ &= m^*(A \cap \bigcup_{i=1}^n E_i) + m^*(A \cap \tilde{E}) \\ &= \sum_{i=1}^n m^*(A \cap E_i) + m^*(A \cap \tilde{E}) \text{ [by theorem 16.9]} \end{aligned}$$

Since this is true for every n,

$$\begin{aligned} m^*A &\geq \sum_{i=1}^{\infty} m^*(A \cap E_i) + m^*(A \cap \tilde{E}) \\ &= m^*(A \cap E) + m^*(A \cap \tilde{E}) \\ &\quad \text{[ since by theorem 16.3 , } m^*(A \cap E) = \sum_{i=1}^{\infty} m^*(A \cap E_i) \text{ ].} \end{aligned}$$

Hence E is measurable.

i.e  $E \in \mathcal{M}$  .

Hence  $\mathcal{M}$  is a  $\sigma$ -algebra.

Hence the theorem.

### Remark:

The intersection of a countable collection of measurable sets is measurable.

Proof:

Since the complement of a measurable set is measurable,

$$\left( \bigcup_{i=1}^{\infty} E_i \right)^c \text{ is measurable.}$$

Hence ,

$(\bigcup_{i=1}^{\infty} \tilde{E}_i)^c$  is measurable.

i.e  $\bigcap_{i=1}^{\infty} E_i$  is measurable.

Hence the required result.

**Theorem 16.11**

The interval  $(a, b)$  is measurable.

**Proof**

Let  $A$  be any subset of  $\mathbb{R}$ .

Let  $A_1 = A \cap (a, b)$   
 $A_2 = A \cap (b, \infty)$   
 $= A \cap (-\infty, b]$ .

Therefore to show that  $(a, b)$  is measurable, we have to prove that

$$m^*A = m^*A_1 + m^*A_2 .$$

If  $m^*A = 0$ , then there is nothing to prove.

Hence let  $m^*A > 0$ .

By definition,

$$m^*A = \inf_{A \subset \bigcup I_n} \sum l(I_n) .$$

Hence for given  $\epsilon > 0$ ,

there exist open intervals  $\{I_n\}_{n=1}^{\infty}$  such that  $A \subset \bigcup_{n=1}^{\infty} I_n$  and

$$\sum_{n=1}^{\infty} l(I_n) = m^*A + \epsilon . \tag{1}$$

Let  $I_n' = I_n \cap (a, b)$

And  $I_n'' = I_n \cap (b, \infty)$ .

Then  $I_n'$  and  $I_n''$  are intervals (or may be empty)

$$\begin{aligned} \text{And } l(I_n) &= l(I_n') + l(I_n'') \\ &= m^*I_n' + m^*I_n'' . \end{aligned} \tag{2}$$

$$\begin{aligned} \text{Now } A_1 &= A \cap (a, b) \\ &\subset \left( \bigcup_{n=1}^{\infty} I_n \right) \cap (a, b) \\ &= \bigcup_{n=1}^{\infty} I_n' \cap (a, b) \\ &= \bigcup_{n=1}^{\infty} I_n' . \end{aligned}$$

$$\text{i.e } A_1 \subset \bigcup_{n=1}^{\infty} I_n' .$$

Therefore,

$$m^*A_1 \subset m^*\left(\bigcup_{n=1}^{\infty} I_n'\right) = \sum_{n=1}^{\infty} m^*I_n'$$

Similarly,

$$m^*A_2 \subset m^*\left(\bigcup_{n=1}^{\infty} I_n''\right) = \sum_{n=1}^{\infty} m^* I_n''.$$

Therefore,

$$\begin{aligned} m^*A_1 + m^*A_2 &= \sum_{n=1}^{\infty} m^* I_n' + \sum_{n=1}^{\infty} m^* I_n'' \\ &= \sum_{n=1}^{\infty} (m^* I_n' + m^* I_n'') \\ &= \sum_{n=1}^{\infty} l(I_n) && \text{[From (2)]} \\ &= m^*A && \text{[From(1) ].} \end{aligned}$$

Since  $A$  was arbitrary,

$$m^*A_1 + m^*A_2 = m^*A.$$

Hence  $(a, \infty)$  is measurable.

Hence the theorem.

**Definition: Borel set**

The smallest  $\sigma$ -algebra which contains all open sets of  $\mathbb{R}$  is called the collection of all Borel sets.

**Theorem 16.12**

Every Borel set is measurable.

**Proof**

To show that every Borel set is measurable, it is enough to show that  $\mathcal{M}$  contains all open sets.

i.e every open set of  $\mathbb{R}$  is measurable.

Then  $\mathcal{M}$  being a  $\sigma$ -algebra and

$\mathcal{P}$  being the smallest  $\sigma$ -algebra containing the open sets of  $\mathbb{R}$ ,

$$\mathcal{P} \subset \mathcal{M}.$$

And hence every element of  $\mathcal{P}$  is measurable

i.e every Borel set is measurable.

Claim: Every open set of  $\mathbb{R}$  is measurable.

Consider  $(a, \infty)$ , for any  $a \in \mathbb{R}$ .

By theorem 16.11,

$(a, \infty)$  is measurable.

Therefore,

$$(a, \infty) \sim (-\infty, a] \text{ is measurable.} \tag{1}$$

For any  $b \in \mathbb{R}$ ,

$$(-\infty, b) = \bigcup_{n=1}^{\infty} (-\infty, b - \frac{1}{n}].$$

By (1),

each  $(-\infty, b - \frac{1}{n}]$  is measurable.

Also union of countable number of measurable sets is measurable.

Hence  $(-\infty, b)$  is measurable.

Now consider any open interval  $(a, b)$ .

$(a, b)$  can be written as

$$(a, b) = (-\infty, b) \cap (a, \infty).$$

As we have already seen,

$(-\infty, b)$  and  $(a, \infty)$  are measurable and  
intersection of two measurable sets is measurable.

Hence  $(a, b)$  is measurable.

Since each open set of  $\mathbb{R}$  is a countable union of open intervals,  
each open set of  $\mathbb{R}$  is measurable.

Hence our claim is proved.

Hence the theorem.

## 16.4 Lebesgue measure

### Definition

The Lebesgue measure  $m$  is the set function from the family  $\mathcal{M}$  of Lebesgue measurable sets to  $\mathbb{R} \cup \{ \infty \}$  defined by

$$m(E) = m^*(E), \text{ for every } E \in \mathcal{M}.$$

i.e.  $m: \mathcal{M} \rightarrow \mathbb{R} \cup \{ \infty \}$  defined by  $m(E) = m^*(E)$ , for every  $E \in \mathcal{M}$ .

### Theorem 16.13

Let  $\{ E_i \}$  be a sequence of pairwise disjoint measurable sets.

Then  $m\left(\bigcup_{i=1}^{\infty} E_i\right) = \sum_{i=1}^{\infty} m(E_i)$ .

### Proof

Let  $E_1, E_2, \dots, E_n$  be a finite sequence of measurable sets.

Then by theorem 16.9,

$$m^*\left(A \cap \left[\bigcup_{i=1}^n E_i\right]\right) = \sum_{i=1}^n m^*(A \cap E_i). \quad \dots(1)$$

where  $A$  is any subset of  $\mathbb{R}$ .

Taking  $A=\mathbb{R}$  in (1), we get

$$m^*\left(\mathbb{R} \cap \left[\bigcup_{i=1}^n E_i\right]\right) = \sum_{i=1}^n m^*(\mathbb{R} \cap E_i).$$

Hence,

$$m^*\left(\bigcup_{i=1}^n E_i\right) = \sum_{i=1}^n m^* E_i.$$

Hence,

$$m\left(\bigcup_{i=1}^n E_i\right) = \sum_{i=1}^n m E_i. \quad \dots(2)$$

Thus  $m$  is finitely additive..

Let  $\{ E_i \}$  be an infinite sequence of positive disjoint measurable sets.

Then,

$$\bigcup_{i=1}^{\infty} E_i \supset \bigcup_{i=1}^n E_i, \text{ for any } n.$$

Therefore,

$$m^*(\bigcup_{i=1}^{\infty} E_i) = m^*(\bigcup_{i=1}^n E_i), \text{ for any } n.$$

Therefore,

$$m(\bigcup_{i=1}^{\infty} E_i) = m(\bigcup_{i=1}^n E_i), \text{ for any } n.$$

Therefore,

$$m(\bigcup_{i=1}^{\infty} E_i) = \sum_{i=1}^n mE_i, \text{ for any } n. \quad [\text{using (2)}]$$

Therefore,

$$m(\bigcup_{i=1}^{\infty} E_i) = \sum_{i=1}^{\infty} m(E_i). \quad \dots(3)$$

Since  $m^*$  is countably subadditive ,  
 $m$  is also countably subadditive on  $\mathbb{M}$  .

Therefore,

$$m(\bigcup_{i=1}^{\infty} E_i) = \sum_{i=1}^{\infty} m(E_i). \quad \dots(4)$$

From (3) and (4),

$$m(\bigcup_{i=1}^{\infty} E_i) = \sum_{i=1}^{\infty} m(E_i).$$

Hence the theorem.

**Theorem 16.14**

Let  $\{E_n\}$  be an infinite decreasing sequence of measurable sets.  
 i.e  $E_{n+1} \subset E_n$  for each  $n$ .  
 Let  $mE_1$  be finite.

Then  $m(\bigcap_{i=1}^{\infty} E_i) = \lim_{n \rightarrow \infty} mE_n$ .

**Proof**

Let  $E = \bigcap_{i=1}^{\infty} E_i$  .

Let  $F_i = E_i - E_{i+1}$  .

Claim: (i)  $E_1 - E = \bigcup_{i=1}^{\infty} F_i$  and

(ii)  $F_i$ 's are pairwise disjoint.

Let  $x \in E_1 - E$ .

Then  $x \in E_1$

and  $x \notin E = \bigcap_{i=1}^{\infty} E_i$  .

i.e  $x \in E_1$  and  $x \notin E_i$  for some  $i$ .

Since  $\{E_n\}$  is an infinite decreasing sequence of measurable sets,  
 $E_1 \supset E_2 \supset \dots E_i$ , for all  $i$ .

Let  $j$  be the smallest suffix for which  $x \notin E_j$ .

i.e  $x \in E_{j-1}$  but  $x \notin E_j$ .

Therefore,

$$x \in E_{j-1} - E_j = F_{j-1}.$$

Therefore,

$$x \in \bigcup_{i=1}^{\infty} F_i .$$

Hence,

$$E_1 - E \subset \bigcup_{i=1}^{\infty} F_i . \quad \dots(1)$$

Let  $y \in \bigcup_{i=1}^{\infty} F_i .$

This implies,

$$y \in F_i \text{ for some } i.$$

This implies,

$$y \in E_i \text{ but } y \notin E_{i+1}.$$

Since  $E_1 \supset E_i$ ,

$$y \in E_1.$$

Since  $y \notin E_{i+1}$ .

$$y \notin \bigcap_{i=1}^{\infty} E_i = E.$$

Therefore ,

$$y \in E_1 - E.$$

Hence,

$$\bigcup_{i=1}^{\infty} F_i \subset E_1 - E. \quad \dots(2)$$

From (1) and (2),

$$E_1 - E = \bigcup_{i=1}^{\infty} F_i .$$

Hence (i) is proved.

To prove (ii) of the claim

i.e  $F_i$ 's are pairwise disjoint, where  $F_i = E_i - E_{i+1}$ .

Let  $i < j$ .

Without loss of generality, assume that  $i < j$ .

Then  $E_i \supset E_j$ .

Let  $x \in F_i$ .

This implies,

$$x \in E_i \text{ and } x \notin E_{i+1}.$$

Since  $E_{i+1} \supset E_i \supset E_j$ ,

this implies,

$$x \notin E_j$$

Therefore,

$$x \notin F_j.$$

Therefore,

$$F_i \cap F_j = \emptyset.$$

Similarly we can prove that  $F_j \not\subset F_i$ .  
Hence the  $F_i$ 's are pairwise disjoint.  
Hence our claim is proved.

Now let us prove the theorem.

Since  $E_i$ , and  $E_{i+1}$  are measurable,  
 $F_i = E_i - E_{i+1}$  is measurable, for all  $i$ .

Also  $E_1 - E = \bigcup_{i=1}^{\infty} F_i$  and

$$F_i \cap F_j = \emptyset, \text{ for } i \neq j.$$

Therefore ,

$$\begin{aligned} m(E_1 - E) &= m\left(\bigcup_{i=1}^{\infty} F_i\right) \\ &= \sum_{i=1}^{\infty} m(F_i) \\ &= \sum_{i=1}^{\infty} m(E_i - E_{i+1}) \end{aligned} \quad \dots(3)$$

Since  $E_1 \supset E$ ,

$$E_1 = E \cup (E_1 - E) \text{ is a disjoint union.}$$

Since  $E_1$  and  $E$  are measurable,

$E_1 - E$  is measurable.

Therefore,

$$m(E_1) = m(E) + m(E_1 - E) \quad \dots(4)$$

Similarly ,

since  $E_i \supset E_{i+1}$ ,

$$m(E_i) = m(E_{i+1}) + m(E_i - E_{i+1}) \quad \dots(5)$$

Since  $E_i \subset E_1$ , for all  $i$ ,

$$m(E_i) \leq m(E_1), \text{ for all } i.$$

Given  $m(E_1) < \infty$ .

Therefore ,

$$m(E_i) < \infty, \text{ for all } i.$$

From(4),

$$m(E_1 - E) = m(E_1) - m(E) \quad \dots(6)$$

from (5),

$$m(E_i - E_{i+1}) = m(E_i) - m(E_{i+1}) \quad \dots(7)$$

Therefore,

$$\begin{aligned} m(E_1) - m(E) &= m(E_1 - E) \\ &= \sum_{i=1}^{\infty} m(E_i - E_{i+1}) \quad \text{[from(3)]} \\ &= \sum_{i=1}^{\infty} (m(E_i) - m(E_{i+1})) \quad \text{[from (7)]} \\ &= \lim_{n \rightarrow \infty} \sum_{i=1}^{n-1} (m(E_i) - m(E_{i+1})) \\ &= \lim_{n \rightarrow \infty} (mE_1 - mE_n) \\ &= mE_1 - \lim_{n \rightarrow \infty} mE_n. \end{aligned}$$

Hence,

$$m(E_1) - m(E) = mE_1 - \lim_{n \rightarrow \infty} mE_n.$$

Since  $m(E_1) < \infty$ , canceling it both sides, we get

$$m(E) = \lim_{n \rightarrow \infty} mE_n.$$

$$\text{i.e. } m\left(\bigcap_{i=1}^{\infty} E_i\right) = \lim_{n \rightarrow \infty} mE_n.$$

Hence the theorem.

**Theorem 16.15 Littlewood's First Principle**

Let E be a given set.

Then the following statements are equivalent:

- (i) E is measurable
- (ii) Given  $\epsilon > 0$ , there is an open set  $U \supset E$  with  $m^*(U - E) < \epsilon$ .
- (iii) Given  $\epsilon > 0$ , there is a closed set  $F \subset E$  with  $m^*(E - F) < \epsilon$ .
- (iv) There is a G set G with  $E \subset G$  such that  $m^*(G - E) = 0$ .
- (v) There is a F set F with  $F \subset E$  such that  $m^*(E - F) = 0$ .

If  $m^*E$  is finite, the above statements are equivalent to

- (vi) Given  $\epsilon > 0$ , there is a finite union V of open intervals such that  $m^*(V - E) < \epsilon$ .

**Proof**

To prove that (i) implies (ii).

Let E be measurable.

Case.1: Suppose  $m^*E = mE < \infty$ .

Then given  $\epsilon > 0$ , there exist open intervals  $I_n, n=1,2,3,\dots$

Such that  $E \subset \bigcup_{n=1}^{\infty} I_n$  and

$$\text{(or) } l(I_n) = m^*E + \epsilon \quad \dots(1)$$

Let  $U = \bigcup_{n=1}^{\infty} I_n$ .

Then U is an open set, [since each  $I_n$  is open.]

$U \supset E$  and

U is measurable. [ since each  $I_n$  is measurable]

By the countable subadditivity of m,

$$\begin{aligned} m^*U &= m\left(\bigcup_{n=1}^{\infty} I_n\right) \\ &= \sum_{n=1}^{\infty} m(I_n) \\ &= \sum_{n=1}^{\infty} l(I_n) \end{aligned} \quad \dots(2)$$

From (1) and (2),

$$m^*E + \epsilon = m^*U \quad \dots(3)$$

Also  $U = E \cup (U - E)$  is a disjoint union.

Therefore,

$$m^*U = m^*E + m^*(U - E)$$

Therefore ,

$$m^*(U - E) = m^*U - m^*E$$

[by (3)]

Hence there is an open set  $U \supset E$  with  $m^*(U - E) < \epsilon$ .

Case .2 Suppose  $m^*E = mE = \epsilon$ .

Let  $X = \bigcup_{n=0}^{\infty} I_n$  where  $I_n = [n, n+1)$

$Y = \bigcup_{n=0}^{\infty} I_n'$  where  $I_n' = (-n-1, -n]$ .

Then  $R = X \cup Y$ .

Now,  $I_n, I_n'$  are countable union of disjoint finite intervals.

Rename their union as  $\{J_n\}_{n=1}^{\infty}$ .

Hence  $E = E \cap R$

$$= E \cap \bigcup_{n=1}^{\infty} J_n$$

$$= \bigcup_{n=1}^{\infty} (E \cap J_n)$$

$$= \bigcup_{n=1}^{\infty} E_n, \quad \text{where } E_n = E \cap J_n.$$

Since  $E$  and  $J_n$  are measurable,  $E_n$ 's are also measurable.

Also  $E_n \subset J_n$ .

Hence  $mE_n = mJ_n < \epsilon/2^n$ .

Hence by case.1 ,

we can find open sets  $U_n$  such that  $E_n \subset U_n$   
and  $m(U_n - E_n) < \epsilon/2^n$ .

Let  $U = \bigcup_{n=1}^{\infty} U_n$ .

Then  $U - E = \bigcup_{n=1}^{\infty} U_n - \bigcup_{n=1}^{\infty} E_n$   
 $\subset \bigcup_{n=1}^{\infty} (U_n - E_n)$

Therefore,

$$\begin{aligned} m(U - E) &= m\left(\bigcup_{n=1}^{\infty} (U_n - E_n)\right) \\ &= \sum_{n=1}^{\infty} m(U_n - E_n) \\ &< \sum_{n=1}^{\infty} \epsilon / 2^n \\ &= \epsilon. \end{aligned}$$

Hence,

$$m^*(U - E) < \epsilon.$$

From case .1 and 2,  
we get that (i) implies (ii).

To Prove that (ii) implies (iv):

Assume that for every  $\epsilon > 0$ ,  
there exists an open set  $U \supset E$  with  $m^*(U - E) < \epsilon$ .

Hence for  $\epsilon = 1/n, n=1,2,3,\dots$ ,  
there exist open sets  $U_n \supset E$  with  $m^*(U_n - E) < 1/n$ .

Let  $G = \bigcap_{n=1}^{\infty} U_n$ .

Then  $G \supset E$  and

Since  $G$  is a countable union of open sets,  
 $G$  is a  $G$  set.

Since  $G \subset U_n$ , for every  $n$ ,  
 $G - E \subset U_n - E$ , for every  $n$ .

Therefore,  
 $m^*(G - E) = m^*(U_n - E) < 1/n$ , for every  $n$ .

Therefore,  
 $m^*(G - E) = 0$ .

Hence,  
there is a  $G$  set  $G$  with  $E \subset G$  such that  $m^*(G - E) = 0$ .

Hence (ii) implies (iv).

To prove that (iv) implies (i):

Assume that there is a  $G$  set  $G$  with  $E \subset G$  such that  $m^*(G - E) = 0$ .

Since any  $G$  set is measurable,  
 $G$  is measurable.

Since  $m^*(G - E) = 0$ ,  
 $G - E$  is measurable.

Hence  $E = G - (G - E)$  is measurable.

Hence (iv) implies (i).

Hence it follows that,  
(i)  $\Rightarrow$  (ii)  $\Rightarrow$  (iv)  $\Rightarrow$  (i).

Hence we get that,  
(i)  $\Leftrightarrow$  (ii)  $\Leftrightarrow$  (iv). ...(1)

To prove that (ii) implies (iii):

Assume (ii)  
i.e Given  $\epsilon > 0$ , there is an open set  $U \supset E$  with  $m^*(U - E) < \epsilon$ .

From(1),  
(ii) implies (i).

Hence  $E$  is measurable.

Hence  $\tilde{E}$  is measurable.

Hence for given given  $\epsilon > 0$ ,  
there is an open set  $U \supset \tilde{E}$  with  $m^*(U - \tilde{E}) < \epsilon$ . [Since (i) implies (ii).]

Since  $U \supset \tilde{E}$ ,

$$\begin{aligned} \tilde{U} &\subset E \text{ and} \\ U - \tilde{E} &= E - \tilde{U} \\ &= U \cap E. \end{aligned}$$

Since  $U$  is open,

$\tilde{U}$  is closed.

Also  $m^*(E - \tilde{U}) = m^*(U - \tilde{E}) < \epsilon$ .

Hence,

there is a closed set  $\tilde{U} \subset E$  with  $m^*(E - \tilde{U}) < \epsilon$ .

Hence (ii) implies (iii)

To prove that (iii) implies (v):

Assume (iii)

i.e Given  $\epsilon > 0$ , there is a closed set  $F \subset E$  with  $m^*(E - F) < \epsilon$ .

Take  $\epsilon = 1/n$ .

Then for each  $n$ ,

there is a closed set  $F_n \subset E$  with  $m^*(E - F_n) < 1/n$ .

Let  $F = \bigcup_{n=1}^{\infty} F_n$ .

Since  $F$  is a countable union of closed sets,

$F$  is a  $F_\sigma$  set.

Also  $F_n \subset E$ , for every  $n$ .

Hence,

$$F = \bigcup_{n=1}^{\infty} F_n \subset E \text{ and}$$

$$E - F \subset E - F_n, \text{ for every } n.$$

Hence,

$$m^*(E - F) = m^*(E - F_n) < 1/n, \text{ for every } n.$$

Hence,

$$m^*(E - F) = 0.$$

Hence,

there is a  $F_\sigma$  set  $F$  with  $F \subset E$  such that  $m^*(E - F) = 0$ .

Hence (iii) implies (v)

To prove that (v) implies (i).

Assume (v)

i.e There is a  $F_\sigma$  set  $F$  with  $F \subset E$  such that  $m^*(E - F) = 0$ .

Since each  $F_n$  set is measurable,

$F$  is measurable.

Since  $m^*(E - F) = 0$ ,

$E - F$  is measurable.

Hence,

$$E = F \cup (E - F) \text{ is measurable.}$$

Hence (V) implies (i).

Hence we have shown that,

$$(ii) \Rightarrow (iii) \Rightarrow (v) \Rightarrow (i)$$

...(2)

From (1) and (2),

it is clear that (i) to (v) are equivalent.  
Hence the theorem.

### Check your progress

1. Prove that properties (i) to (v) are equivalent to (vi), if  $m^*E$  is finite.
2. Show that if  $E$  is measurable, then each translate  $E+y$  is also measurable.
3. Show that if  $E_1$  and  $E_2$  are measurable, then  $m(E_1 \cup E_2) + m(E_1 \cap E_2) = mE_1 + mE_2$ .
4. Let  $\{E_i\}$  be a sequence of disjoint measurable sets and  $A$  be any set.

$$\text{Show that } m^*(A \cap \bigcup_{i=1}^{\infty} E_i) = \sum_{i=1}^{\infty} m^*(A \cap E_i)$$

### 16.5 Let us sum up

Thus in this lesson, we have seen

- The definition of outer measure of sets
- Outer measure of an interval is its length
- Some important properties of Outer measure
- The definition of Measurable sets
- Countable union of measurable sets is also measurable
- Countable intersection of measurable sets is also measurable
- Every Borel set is measurable and
- Littlewood's First Principle.

### 16.6 Lesson End Activities

1. Prove that  $m^*$  is translation invariant
2. If  $E_1$  and  $E_2$  are measurable sets, Prove that  $m(E_1 \cup E_2) + m(E_1 \cap E_2) = m(E_1) + m(E_2)$
3. Prove that  $m^*(A) = 0$  if  $A$  is a countable set.

### 16.7 References

1. Real Analysis by H.L. Royden (third Edition)
2. Measure theory and Intergration by G. deBarra.

## LESSON – 17

### MEASURABLE FUNCTIONS AND LITTLEWOOD'S THEOREM

#### Contents

#### 17.0 Introduction

#### 17.1 Aims and objectives

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#### 17.6 References

#### 17.0 Introduction

Already we have studied Littlewood's first principle. In this lesson we are going to study about measurable functions and littlewood's second and third principles.

#### 17.1 Aims and objectives

After studying this lesson, you would know

- What are measurable functions?
- Sum, difference, scalar product and product of measurable functions are measurable and
- Littlewood's Theorems

#### 17.2 Measurable functions

##### Theorem 17.1

Let  $f$  be an extended real valued function whose domain is measurable.

Then the following statements are equivalent:

- (i) For each real number  $a$ , the set  $\{x \in D : f(x) > a\}$  is measurable.
- (ii) For each real number  $a$ , the set  $\{x \in D : f(x) \leq a\}$  is measurable.
- (iii) For each real number  $a$ , the set  $\{x \in D : f(x) < a\}$  is measurable.
- (iv) For each real number  $a$ , the set  $\{x \in D : f(x) \geq a\}$  is measurable.

These statements imply

- (v) For each real number  $a$ , the set  $\{x \in D : f(x) = a\}$  is measurable.

#### Proof

Let the domain of  $f$  be  $D$ .

Then  $D$  is measurable.

To prove that (i) implies (iv)

Let  $\{x \in D : f(x) > a\}$  is measurable, for every real  $a$ .

Since  $\{x \in D : f(x) \leq a\} = D - \{x \in D : f(x) > a\}$  and difference of two measurable sets is measurable,

$\{x \in D : f(x) \leq a\}$  is measurable.

Hence (i) implies (iv)

To prove that (iv) implies (i)

Assume (iv). i.e  $\{x \in D : f(x) \geq a\}$  is measurable, for every real  $a$ .

Then  $\{x \in D : f(x) > a\} = D - \{x \in D : f(x) \leq a\}$

being the difference of two measurable sets is measurable.  
Hence (iv) implies (i)

Hence (i)  $\Leftrightarrow$  (iv) (1)

To prove that (ii) implies (iii)

Assume(ii) i.e  $\{x / f(x) > \alpha\}$  is measurable, for every real  $\alpha$ .

Then  $\{x / f(x) < \alpha\} = D - \{x / f(x) > \alpha\}$

being the difference of two measurable sets is measurable.

Hence (ii) implies (iv)

To prove that (iii) implies (ii)

Assume (iii) i.e  $\{x / f(x) < \alpha\}$  is measurable, for every real  $\alpha$ .

Then  $\{x / f(x) > \alpha\} = D - \{x / f(x) < \alpha\}$

being the difference of two measurable sets is measurable.

Hence (iii) implies (ii)

Hence (ii)  $\Leftrightarrow$  (iii) ... (2)

To prove that (i) implies (ii)

Assume (i) i.e  $\{x / f(x) > \alpha\}$  is measurable, for every real  $\alpha$ .

Therefore  $\{x / f(x) > -1/n\}$  is measurable, for  $n=1,2,3,..$

Also countable intersection of measurable sets is measurable.

Hence ,

$$\{x / f(x) > \alpha\} = \bigcap_{n=1}^{\infty} \{x / f(x) > \alpha - \frac{1}{n}\} \text{ is measurable.}$$

Hence (i) implies (ii)

To prove that (ii) implies (i)

Assume (ii) i.e  $\{x / f(x) < \alpha\}$  is measurable, for every real  $\alpha$ .

Then for  $n=1, 2, 3, \dots,$

$\{x / f(x) < \alpha + 1/n\}$  is measurable.

Also countable union of measurable sets is measurable.

Hence,

$$\{x / f(x) > \alpha\} = \bigcup_{n=1}^{\infty} \{f(x) \geq \alpha + \frac{1}{n}\} \text{ is measurable.}$$

Hence (ii) implies (i).

Hence (i)  $\Leftrightarrow$  (ii) ... (3)

From(1), (2) and (3),

$$(iv) \Leftrightarrow (i) \Leftrightarrow (ii) \Leftrightarrow (iii)$$

Hence the first four statements are equivalent.

To prove (v):

Case1:  $c = \infty$ .

Then  $\{x / f(x) = c\} = \{x / f(x) < \infty\} \cap \{x / f(x) > -\infty\}$ .

Since intersection of two measurable sets is measurable,  
 $\{x / f(x) = c\}$  is measurable.

Case2:  $c = -\infty$ .

Then  $\{x / f(x) = -\infty\} = \{x / f(x) = -n\}$   
 $= \bigcap_{n=1}^{\infty} \{x / f(x) \geq n\}$ .

By assumptions  $\{x / f(x) \geq n\}$  is measurable for every  $n=1,2,3,\dots$

Also countable intersection of measurable sets is measurable.

Hence  $\{x / f(x) = -\infty\}$  is measurable.

Case3.  $c = -$ .

Then  $\{x / f(x) = -\infty\} = \bigcap_{n=1}^{\infty} \{x / f(x) \leq -n\}$

By assumptions  $\{x / f(x) \leq -n\}$  is measurable, for  $n=1,2,3,\dots$

Also countable intersection of measurable sets is measurable.

Hence  $\{x / f(x) = -\infty\}$  is measurable.

Hence the set  $\{x / f(x) = c\}$  is measurable, for each real  $c$ .

Hence (v) follows from (i),(ii),(iii) and (iv).

Hence the theorem.

**Definition**

An extended real valued function  $f$  is Lebesgue measurable if its domain is measurable and if it satisfies one of the following four statements:

- (i) For each real number  $c$ , the set  $\{x / f(x) > c\}$  is measurable.
- (ii) For each real number  $c$ , the set  $\{x / f(x) < c\}$  is measurable.
- (iii) For each real number  $c$ , the set  $\{x / f(x) < c\}$  is measurable.
- (iv) For each real number  $c$ , the set  $\{x / f(x) < c\}$  is measurable.

**Check your progress**

1. Show that constant functions are measurable.
2. Show that if  $A$  is measurable subset of  $R$ , then its characteristic function  $\chi_A$  is a measurable function.
3. Show that continuous functions from  $R \rightarrow R$  are measurable.
4. Show that any step function is measurable.

**Answers**

1. Let  $f : D \rightarrow R \cup \{-\infty, \infty\}$  be defined as  
 $f(x) = c$ , for every  $x \in D$ ,  
 where  $D$  is a measurable subset of  $R$ .

Now If  $c < \infty$ ,  
 then  $\{x / f(x) > c\} = D$

And If  $c > -\infty$ ,  
 then  $\{x / f(x) > c\} = \emptyset$ .

Since both  $D$  and  $\chi_A$  are measurable functions,  
 $f$  is also a measurable function.  
Hence constant functions are measurable.

$$2. \quad \chi_A(x) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A \end{cases}$$

Therefore,

$$\{x / \chi_A(x) > \alpha\} = \begin{cases} R & \text{if } \alpha < 0 \\ A & \text{if } 0 \leq \alpha < 1 \\ \emptyset & \text{if } \alpha \geq 1 \end{cases}$$

Since  $R$ ,  $A$ , and  $\emptyset$  are measurable functions,  
 $\chi_A$  is also a measurable function.

3. Assume that  $f$  is a continuous function from  $R$  to  $R$ .

$$\text{Then } \{x / f(x) > \alpha\} = f^{-1}((\alpha, \infty))$$

Since  $(\alpha, \infty)$  is open in  $R$  and  $f$  is continuous,  
 $f^{-1}((\alpha, \infty))$  is open in  $R$ .

Since any open set in  $R$  is measurable,  
 $f^{-1}((\alpha, \infty))$  is also measurable.

Hence  $\{x / f(x) > \alpha\}$  is measurable.

Hence  $f$  is measurable.

Hence continuous functions from  $R \rightarrow R$  are measurable.

4. A real valued function  $f$  defined on  $[a, b]$  is called a step function if there is a partition

$$a = x_0 < x_1 < \dots < x_n = b,$$

such that for each  $i$ , the function assumes only one value in the interval  $(x_{i-1}, x_i)$ .

$$\text{Suppose if } f(x) = \alpha_{i-1}, i=1,2,\dots,n.$$

$$\text{Then } \{x / f(x) > \alpha\} = \begin{cases} \emptyset & \text{if } \alpha \geq \alpha_{i-1}, i=1,2,\dots,n \\ [a, b] & \text{if } \alpha < \alpha_{i-1}, i=1,2,\dots,n \\ \text{a finite union of open intervals} & \text{if } \alpha \text{ lies between min and max of } \{\alpha_0, \alpha_1, \dots, \alpha_n\}. \end{cases}$$

since  $\emptyset$ ,  $[a, b]$  and finite union of open intervals are all measurable,  
 $\{x / f(x) > \alpha\}$  is also measurable.

**Theorem 17.2**

Let  $c$  be a constant and  $f$  and  $g$  are two measurable real valued functions defined on the same domain. Then the functions  $f+c$ ,  $cf$ ,  $f+g$ ,  $f-g$  and  $fg$  are also measurable.

**Proof:**

(i) To show that  $f+c$  is measurable.

$$\text{Consider } \{x / f(x)+c > \alpha\} = \{x / f(x) > \alpha - c\}.$$

Since  $f$  is measurable,

$$\{x / f(x) > \alpha - c\} \text{ is measurable.}$$

Hence,

$$\{x / f(x)+c > \alpha\} \text{ is measurable.}$$

Hence  $f+c$  is measurable.

(ii) To show that  $cf$  is measurable:

Case1:  $c>0$

$$\begin{aligned} \text{Consider } \{x / (cf)(x) > \alpha\} &= \{x / cf(x) > \alpha\} \\ &= \{x / f(x) > \alpha / c\} \end{aligned}$$

Since  $f$  is measurable,

$$\{x / f(x) > \alpha / c\} \text{ is measurable.}$$

Hence  $\{x / (cf)(x) > \alpha\}$  is measurable.

Hence  $cf$  is measurable, for all real constants  $c$ .

Case2:  $c=0$ .

Then  $cf=0$  is a constant function.

Hence  $cf$  is measurable.

Case3:  $c<0$ .

$$\text{Then } \{x / (cf)(x) > \alpha\} = \{x / f(x) < \alpha / c\}.$$

Since  $f$  is measurable,

$$\{x / f(x) < \alpha / c\} \text{ is measurable.}$$

Hence  $\{x / (cf)(x) > \alpha\}$  is measurable.

Hence  $cf$  is measurable.

(iii) To show that  $f+g$  is measurable.

$$\begin{aligned} \text{Consider } \{x / (f+g)(x) < \alpha\} &= \{x / f(x) + g(x) < \alpha\} \\ &= \{x / f(x) < \alpha - g(x)\}. \end{aligned}$$

Since between any two reals, there is a rational,

there exists a rational  $r$  such that  $f(x) < r < \alpha - g(x)$ .

Therefore,

$$\{x / (f+g)(x) < \alpha\} = \bigcup_r [\{x / f(x) < r\} \cap \{x / g(x) < \alpha - r\}].$$

Since  $f$  and  $g$  are measurable,

$$\{x / f(x) < r\} \text{ and } \{x / g(x) < \alpha - r\} \text{ are measurable.}$$

Hence,

$$\{x / f(x) < r\} \cap \{x / g(x) < \alpha - r\} \text{ is measurable.}$$

Since the rationals are countable and

countable union of measurable sets is measurable,

$$\{x / (f+g)(x) < \alpha\} \text{ is measurable.}$$

Hence  $f+g$  is measurable.

(iv) To show that  $f-g$  is measurable.

$$\text{Now } f-g = f+(-1)g.$$

From (ii),

$$(-1)g \text{ is measurable.}$$

Hence from (iii),

$f + (-1)g$  is measurable.

Hence  $f - g$  is measurable.

(v) To show that  $f^2$  is measurable.

Consider  $\{x / f^2(x) > \alpha\}$

$$= \{x / f(x) > \sqrt{\alpha}\} \cup \{x / f(x) < -\sqrt{\alpha}\} \quad \text{when } \alpha \geq 0.$$

$$= \emptyset, \quad \text{when } \alpha < 0.$$

Since  $f$  is measurable,

$\{x / f(x) > \sqrt{\alpha}\}$  and  $\{x / f(x) < -\sqrt{\alpha}\}$  are measurable.

Hence,

$\{x / f(x) > \sqrt{\alpha}\} \cup \{x / f(x) < -\sqrt{\alpha}\}$  is measurable.

Since  $D$  is the domain of  $f$ ,

$D$  is also measurable.

Hence  $\{x / f^2(x) > \alpha\}$  is measurable.

Hence  $f^2$  is measurable.

(vi) To show that  $fg$  is measurable.

From (iii) and (iv),

$f+g$  and  $f-g$  are measurable.

Hence by (v),

$(f+g)^2$  and  $(f-g)^2$  are measurable.

Hence again by (iv),

$[(f+g)^2 - (f-g)^2]$  is measurable.

Hence by (ii),

$(1/4)[(f+g)^2 - (f-g)^2]$  is measurable.

Hence  $fg$  is measurable.

Hence the theorem.

### Theorem 17.3

Let  $f_1, f_2, \dots, f_n$  be measurable functions defined on the same domain  $D$ .

Then  $\max(f_1, f_2, \dots, f_n)$  and  $\min(f_1, f_2, \dots, f_n)$  are measurable.

#### Proof

Let  $h = \max(f_1, f_2, \dots, f_n)$

Therefore,

$$h(x) = \max(f_1(x), f_2(x), \dots, f_n(x)), \text{ for all } x \in D.$$

Therefore,

$$\{x / h(x) > \alpha\} = \bigcup_{i=1}^n \{x / f_i(x) > \alpha\}.$$

Since each  $f_i$  is measurable,

$\{x / f_i(x) > \alpha\}$  is measurable, for  $i=1, 2, \dots, n$ , and for every real  $\alpha$ .

Since finite union of measurable sets is also measurable,

$\{x / h(x) > \alpha\}$  is measurable.

Hence  $h = \max(f_1, f_2, \dots, f_n)$  is measurable.

$$\text{Let } g = \min (f_1, f_2, \dots, f_n)$$

Therefore,

$$g(x) = \min ( f_1(x), f_2(x), \dots, f_n(x) ) , \quad \text{for all } x \in D.$$

Therefore,

$$\{x / g(x) < \alpha\} = \bigcup_{i=1}^n \{x / f_i(x) < \alpha\}$$

Since each  $f_i$  is measurable,

$$\{x / f_i(x) < \alpha\} \text{ is measurable, for } i=1,2,\dots,n, \text{ and for every real } \alpha.$$

Since finite intersection of measurable sets is also measurable,

$$\{x / g(x) < \alpha\} \text{ is measurable.}$$

Hence  $g = \min (f_1, f_2, \dots, f_n)$  is measurable.

Hence the theorem.

Remarks:

From the above theorem, we get the following important results.

1. If  $f$  and  $g$  are measurable, then  $f \vee g = \max(f, g)$  is measurable.
2. If  $f$  and  $g$  are measurable, then  $f \wedge g = \min(f, g)$  is measurable.
3. If  $f$  is measurable,  $f^+ = f \vee 0 = \max(f, 0)$  is measurable.
4.  $f^+$  is called the positive part of  $f$ .
5. If  $f$  is measurable,  $f^- = f \wedge 0 = \min(f, 0)$  is measurable.
6.  $f^-$  is called the negative part of  $f$ .
7. If  $f^+$  and  $f^-$  are measurable, then  $f = f^+ - f^-$  is measurable.  
and  $|f| = f^+ + f^-$  is measurable.
8. From 3,5 and 7, it is clear that if  $f$  is measurable, then  $|f|$  is measurable.
9. But the converse is not true. i.e if  $|f|$  is measurable, then  $f$  need not be measurable.

**Theorem 17.4**

Let  $\{f_n\}$  be a sequence of measurable functions with the same domain of definition.

Then the functions  $\sup_n f_n, \inf_n f_n, \overline{\lim} f_n, \underline{\lim} f_n$  are all measurable.

**Proof**

Let  $\{f_n\}$  be a sequence of measurable functions with the same domain of definition.

To prove that  $\sup_n f_n$  is measurable.

$$\text{Let } g(x) = \sup_n f_n(x).$$

$$\text{Consider } \{x / g(x) > \alpha\} = \bigcup_{n=1}^{\infty} \{x / f_n(x) > \alpha\}.$$

Since each  $f_n$  is measurable,

$$\{x / f_n(x) > \alpha\} \text{ is measurable, for each } n, \text{ and for every real } \alpha.$$

Also countable union of measurable sets is measurable.

Hence  $\{x / g(x) > \alpha\}$  is measurable.

Hence  $g = \sup_n f_n$  is measurable. ...(1)

To prove that  $\inf_n f_n$  is measurable.

Let  $h(x) = \inf_n f_n(x)$ .

Consider  $\{x / h(x) < \alpha\} = \bigcup_{n=1}^{\infty} \{x / f_n(x) < \alpha\}$ .

Since each  $f_n$  is measurable,

$\{x / f_n(x) < \alpha\}$  is measurable, for each  $n$ , and for every real  $\alpha$ .

Also countable intersection of measurable sets is measurable.

Hence  $\{x / h(x) < \alpha\}$  is measurable.

Hence  $h = \inf_n f_n$  is measurable. ...(2)

To prove that  $\overline{\lim} f_n$  is measurable.

Now  $\overline{\lim} f_n = \inf_n (\sup_{k \geq n} f_k) = \inf_n g_n$ .

Since each  $f_k$  is measurable,

$g_n = \sup_{k \geq n} f_k$  is measurable, for every  $n$ . [from(1)]

Therefore  $\inf_n g_n$  is measurable. [from (2)]

Hence  $\overline{\lim} f_n$  is measurable.

To prove that  $\underline{\lim} f_n$  is measurable.

Similarly  $\underline{\lim} f_n = \sup_n (\inf_{k \geq n} f_k) = \sup_n h_n$

Since each  $f_k$  is measurable,

$h_n = \inf_{k \geq n} f_k$  is measurable, for every  $n$ . [from(2)]

Hence  $\sup_n h_n$  is measurable. [from (2)]

Hence  $\underline{\lim} f_n$  is measurable.

Hence the theorem.

### Remarks

.Let  $\{f_n\}$  be a converging sequence of measurable functions on the same domain  $D$ .

Let  $\lim_n f_n = f$ . Then  $f$  is measurable.

### Proof

Since  $\lim_n f_n$  exists,

$$\overline{\lim} f_n = \underline{\lim} f_n = \lim_n f_n.$$

By the above theorem,

both  $\overline{\lim} f_n$  and  $\underline{\lim} f_n$  are measurable.

Hence  $\lim_n f_n$  is measurable.

Hence  $f$  is measurable.

### Definition

A property is said to hold at almost everywhere if the set of points at which the property fails to hold is a set of measure is zero.

Thus, in particular,

$f = g$  a.e if  $f$  and  $g$  have the same domain  $D$  and

$$m\{x / f(x) \neq g(x)\} = 0.$$

**Definition**

If  $\{f_n\}$  is a sequence of functions, then  $\{f_n\}$  is said to converge to  $g$  almost everywhere, if there is a set  $E$  of measure zero such that  $\{f_n(x)\}$  converges to  $g(x)$  for all  $x$  not in  $E$ .

**Theorem 17.5**

If  $f$  is measurable and  $f = g$  a.e., then  $g$  is measurable.

**Proof**

Let  $E = \{x / f(x) \neq g(x)\}$

Since  $f = g$  a.e.,  $mE = 0$ .

Now,

$$\{x / g(x) > \alpha\} = \{x / f(x) > \alpha\} \cup \{x \in E / g(x) > \alpha\} - \{x \in E / g(x) \leq \alpha\}.$$

Since  $\{x \in E / g(x) > \alpha\} \subset E$

and  $\{x \in E / g(x) \leq \alpha\} \subset E$

and  $mE = 0$ , we get that

measures of  $\{x \in E / g(x) > \alpha\}$  and  $\{x \in E / g(x) \leq \alpha\}$  are zero.

Hence,

$\{x \in E / g(x) > \alpha\}$  and  $\{x \in E / g(x) \leq \alpha\}$  are measurable.

Since  $f$  is measurable,

$\{x / f(x) > \alpha\}$  is measurable.

Hence  $\{x / g(x) > \alpha\}$  is measurable, for every real  $\alpha$ .

Hence  $g$  is measurable.

Hence the theorem.

**Theorem 17.6 Littlewood's second principle**

Let  $f$  be a measurable function defined on an interval  $[a, b]$ , and assume that  $f$  takes values  $\pm \infty$  only on a set of measure zero.

Then given  $\epsilon > 0$ , we can find a step function  $g$  and a continuous function  $h$  such that

$$|f - g| < \epsilon \quad \text{and} \quad |f - h| < \epsilon.$$

**Check your progress**

Prove theorem 17.6 using the following results.

- (a) Given a measurable function  $f$  on  $[a, b]$  that takes values  $\pm \infty$  only on a set of measure zero, and given  $\epsilon > 0$ , there is an  $M$  such that  $|f| \leq M$  except on a set of measure less than  $\epsilon/3$ .
- (b) Let  $f$  be a measurable function on  $[a, b]$ . Given  $\epsilon > 0$  and  $M$ , there is a simple function  $m$  such that  $|f(x) - m(x)| < \epsilon$  except where  $|f(x)| > M$ . If  $m \leq M$ , then we may take so that  $m \leq M$ .
- (c) Given a simple function  $m$  on  $[a, b]$ , there is a step function  $g$  on  $[a, b]$  such that  $g(x) = m(x)$  except on a set of measure less than  $\epsilon/3$ . If  $m \leq M$ , we can take  $g$  so that  $m \leq M$ .
- (d) Given a step function  $g$  on  $[a, b]$ , there is a continuous function  $h$  such that  $g(x) = h(x)$  except on a set of measure less than  $\epsilon/3$ . If  $m \leq M$ , we can take  $h$  so that  $m \leq M$ .

**17.3 Littlewood's theorem**

**Theorem 17.7 Littlewood's third principle**

Let  $E$  be a measurable set of finite measure, and  $\{f_n\}$  a sequence of measurable functions defined on  $E$ . Let  $f$  be a real valued function such that for each  $x$  in  $E$  we have  $f_n(x) \rightarrow f(x)$ .

Then given  $\epsilon > 0$  and  $\delta > 0$ , there is a measurable set  $A \subset E$  with  $m(A) < \delta$  and an integer  $N$  such that for all  $x \notin A$  and all  $n \geq N$ ,

$$|f_n(x) - f(x)| < \epsilon.$$

**Proof**

$$\text{Let } G_n = \{x \in E / |f_n(x) - f(x)| < \epsilon\}$$

$$\begin{aligned} \text{Let } E_N &= \bigcup_{n=N}^{\infty} G_n \\ &= \{x \in E / |f_n(x) - f(x)| < \epsilon, \text{ for some } n \geq N\} \end{aligned}$$

$$\begin{aligned} \text{Since } E_{N+1} &= \bigcup_{n=N+1}^{\infty} G_n \\ &\subset \bigcup_{n=N}^{\infty} G_n \\ &= E_N, \end{aligned}$$

$\{E_N\}$  is a decreasing sequence of sets.

Since  $f_n(x) \rightarrow f(x)$ , for every  $x \in E$ ,  
for each  $x \in E$ , there exists some  $E_N$  such that  $x \notin E_N$ .

This implies that,

$$\bigcap E_N = \phi.$$

Since  $f$  and  $f_n$  are measurable functions,  
 $|f_n - f|$  is measurable, for every  $n$ .

This implies that,  
 $|f_n - f|$  is measurable, for every  $n$ .

Therefore,  
 $G_n$  is measurable, for every  $n$ .

Since  $E_N$  is a union of these measurable sets,  
 $E_N$  is measurable, for every  $n$ .

Since  $E$  is of finite measure and  $E_N \subset E$ ,  
 $E_N$ 's are also of finite measure.

Therefore by the theorem 16.14,  
"Let  $\{E_n\}$  be an infinite decreasing sequence of measurable sets.  
i.e  $E_{n+1} \subset E_n$  for each  $n$ . Let  $m(E_1)$  be finite.

$$\text{Then } m\left(\bigcap_{i=1}^{\infty} E_i\right) = \lim_{n \rightarrow \infty} m(E_n)."$$

$$\begin{aligned} \lim_{n \rightarrow \infty} m(E_n) &= m\left(\bigcap_{i=1}^{\infty} E_i\right) \\ &= m(\phi) \\ &= 0. \end{aligned}$$

Hence given  $\epsilon > 0$ , there exists  $N$  such that  
 $m(E_N) < \delta$  and  
 $m(\{x \in E / |f_n(x) - f(x)| < \epsilon, \text{ for some } n \geq N\}) < \delta$ .

Let this  $E_N$  be denoted by  $A$ .

Then  $m(A) < \delta$

And  $\tilde{A} = \{x \in E / |f_n(x) - f(x)| < \epsilon, \text{ for all } n \in \mathbb{N}\}$ .

Hence there is a measurable set  $A \subset E$  with  $m(A) < \epsilon$  and an integer  $N$  such that for all  $x \notin A$  and all  $n \in \mathbb{N}$ ,

$$|f_n(x) - f(x)| < \epsilon.$$

Hence the theorem.

### 17.4 Let us sum up

Thus in this lesson, we have seen

- The definition of measurable function
- Constant functions, Characteristics functions, continuous functions, step functions on a measurable domain are measurable functions.
- Sum, difference, product and scalar multiple of measurable functions are measurable functions.
- $\max(f_1, f_2, \dots, f_n)$  and  $\min(f_1, f_2, \dots, f_n)$  are measurable if  $f_1, f_2, \dots, f_n$  are measurable.
- $\sup_n f_n, \inf_n f_n, \limsup f_n, \liminf f_n$  of a sequence of measurable functions are measurable.
- The definition of almost everywhere and almost everywhere convergence and some theorems based on them
- Littlewood's principles

### 17.5 Lesson End Activities

1.  $D$  is a dense subset of the set of all real numbers. Set  $f$  be an extended real – valued function on  $\mathbb{R}$  such that  $\{x : f(x) > \alpha\}$  is measurable for each  $\alpha \in D$ . prove that  $f$  is a measurable function.

2. if  $\{f_n\}_{n=1}^{\infty}$  is a sequence of measurable functions, prove that the set of points  $x$  at which  $\{f_n(x)\}_{n=1}^{\infty}$  converges is a measurable set.

### 17.6 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## LESSON 18

### THE LEBESGUE INTEGRAL OF BOUNDED FUNCTIONS OVER A SET OF FINITE MEASURE

#### Contents

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#### 18.6 References

#### 18.0 Introduction

In this lesson, we are going to study the Lebesgue integrals of a simple function and bounded functions.

#### 18.1 Aims and objectives

After studying this lesson, you would know

- The definition of a simple function
- The Lebesgue integral of a simple function which vanish outside a set of finite measure.
- Some of its properties
- The Lebesgue integral of bounded functions over a set of finite measure.
- Properties of the Lebesgue integral of bounded functions over a set of finite measure and
- Bounded convergence theorem.

#### 18.2 Lebesgue integral of a simple function which vanish outside a set of finite measure

##### Definition

The function  $\chi_E$  defined by

$$\chi_E(x) = \begin{cases} 1 & \text{if } x \in E \\ 0 & \text{if } x \notin E \end{cases}$$

is called the characteristic function of  $E$ .

##### Definition

A linear combination

$$f(x) = \sum_{i=1}^n a_i \chi_{E_i}(x)$$

is called a simple function if the sets  $E_i$  are measurable.

##### Remark

1. This representation of  $f$  is not unique.
2. A function  $f$  is simple if and only if it is measurable and assumes only a finite number of values.
3. If  $f$  is a simple function and  $\{a_1, a_2, \dots, a_n\}$  the set of non-zero values of  $f$ , then

$$= \sum a_i \chi_{A_i}$$

where  $A_i = \{ x / (x) = a_i \}$ .

This representation of  $f$  is called the canonical representation, where the  $A_i$ 's are disjoint and the  $a_i$ 's are distinct and non-zero.

**Definition**

If  $f$  vanishes outside a set of finite measure and has the canonical representation

$$f = \sum_{i=1}^n a_i \chi_{A_i} ,$$

then we define the integral of  $f$  by

$$\int \varphi(x) dx = \sum_{i=1}^n a_i m A_i .$$

Remark:

4. The expression for the integral of  $f$  is sometimes abbreviated as  $\int \varphi$ .
5. If  $E$  is any measurable set, we define

$$\int_E \varphi = \int \varphi \cdot \chi_E$$

**Theorem 18.1**

Let  $f = \sum a_i \chi_{E_i}$ , with  $E_i \cap E_j = \emptyset$  for  $i \neq j$ .

Suppose each set  $E_i$  is a measurable set of finite measure. Then

$$\int \varphi = \sum_{i=1}^n a_i m E_i .$$

**Proof**

The set  $A_a = \{ x / (x) = a \} = \bigcup_{a_i=a} E_i$ .

Hence  $m A_a = \sum_{a_i=a} a_i m E_i$  (by the additivity of  $m$ )

Therefore ,

$$\begin{aligned} \int \varphi(x) dx &= \sum a m A_a \\ &= \sum_{i=1}^n a_i m E_i . \end{aligned}$$

Hence the theorem.

**Theorem 18.2**

Let  $\varphi$  and  $\chi$  be simple functions which vanish outside a set of finite measure . Then

$$\int (a\varphi + b\chi) = a \int \varphi + b \int \chi ,$$

and if  $\varphi = \chi$  a.e., then

$$\int \varphi = \int \chi .$$

**Proof**

Let  $\{A_i\}$  and  $\{B_i\}$  be the sets occurring in the canonical representation of  $\varphi$  and  $\chi$ .

Let  $A_0$  and  $B_0$  be the sets where  $\varphi$  and  $\chi$  are zero.

Then the sets  $E_k$  obtained by taking intersection  $A_i \cap B_j$  form a finite disjoint collection of measurable sets.

Hence  $\varphi$  and  $\chi$  can be written as

$$\begin{aligned} &= \sum_{k=1}^N a_k \chi_{E_k} , \\ &= \sum_{k=1}^N b_k \chi_{E_k} . \end{aligned}$$

Therefore ,

$$\begin{aligned} a \varphi + b \chi &= \sum_{k=1}^N (aa_k + bb_k) \chi_{E_k} \\ &= a \sum_{k=1}^N a_k \chi_{E_k} + b \sum_{k=1}^N b_k \chi_{E_k} . \end{aligned}$$

Therefore ,

$$\begin{aligned} \int (a\varphi + b\chi) &= a \sum_{k=1}^N a_k mE_k + b \sum_{k=1}^N b_k mE_k . \\ &= a \int \varphi + b \int \chi . \end{aligned}$$

Hence ,

$$\int (a\varphi + b\chi) = a \int \varphi + b \int \chi . \quad \dots(1)$$

If  $\varphi \geq \chi$  a.e.,  
then  $\int \varphi - \int \chi \geq 0$  a.e.

Since the integral of a simple function is always greater than or equal to zero,

$$\int (\varphi - \chi) \geq 0 .$$

From (1)

$$\int (\varphi - \chi) = \int \varphi - \int \chi$$

Therefore,

$$\int \varphi - \int \chi \geq 0 .$$

Hence ,

$$\int \varphi \geq \int \chi .$$

Hence the theorem.

Remark:

From theorem 18.2, if  $\varphi = \sum_{i=1}^n a_i \chi_{E_i}$  , then

$$\int \varphi = \sum_{i=1}^n a_i mE_i .$$

Hence the restriction of theorem 18.1 that the sets  $E_i$  be distinct is unnecessary.

### Theorem 18.3

Let  $f$  be defined and bounded on a measurable set  $E$  with  $mE$  finite. In order that

$$\inf_{f \leq \psi} \int_E \psi(x) dx = \sup_{f \geq \varphi} \int_E \varphi(x) dx$$

for all simple functions  $\psi$  and  $\varphi$ , it is necessary and sufficient that  $f$  be measurable.

**Proof**

Let  $f$  be bounded by  $M$ .

i.e.  $|f(x)| \leq M$ , for all  $x$  in  $E$ .

Assume that  $f$  is measurable.

Then the sets

$$E_k = \left\{ x \mid \frac{kM}{n} \geq f(x) > \frac{(k-1)M}{n} \right\}, \quad -n \leq k \leq n,$$

are measurable, disjoint and

$$E = \bigcup_{k=-n}^n E_k.$$

Therefore  $mE = \sum_{k=-n}^n mE_k$ .

For  $n=1,2,3,\dots$ ,

define the simple functions  $\psi_n$  and  $\varphi_n$  by

$$\psi_n(x) = \frac{M}{n} \sum_{k=-n}^n k \chi_{E_k}(x) \quad \text{and}$$

$$\varphi_n(x) = \frac{M}{n} \sum_{k=-n}^n (k-1) \chi_{E_k}(x)$$

Hence  $\psi_n(x) \leq f(x) < \varphi_n(x)$ ,  $n=1,2,3,\dots$

Therefore

$$\inf_{\psi \leq f} \int_E \psi(x) dx = \int_E \psi_n(x) dx = \frac{M}{n} \sum_{k=-n}^n k mE_k \quad \text{and}$$

$$\sup_{f \geq \varphi} \int_E \varphi(x) dx = \int_E \varphi_n(x) dx = \frac{M}{n} \sum_{k=-n}^n (k-1) mE_k.$$

Hence  $0 \leq \inf_{\psi \leq f} \int_E \psi(x) dx - \sup_{f \geq \varphi} \int_E \varphi(x) dx$

$$\begin{aligned} &= \frac{M}{n} \sum_{k=-n}^n k mE_k - \frac{M}{n} \sum_{k=-n}^n (k-1) mE_k \\ &= \frac{M}{n} \sum_{k=-n}^n mE_k \\ &= \frac{M}{n} mE. \end{aligned}$$

Since  $\lim_{n \rightarrow \infty} \frac{M}{n} mE = 0$ , we get that

$$\inf_{\psi \leq f} \int_E \psi(x) dx = \sup_{f \geq \varphi} \int_E \varphi(x) dx.$$

Hence the condition is sufficient.

To prove the necessary part,

Assume that  $\inf_{\psi \leq f} \int_E \psi(x) dx = \sup_{f \geq \varphi} \int_E \varphi(x) dx$ .

Then given  $n$ ,

there are simple functions  $\phi_n$  and  $\psi_n$  such that

$$\phi_n(x) \leq f(x) \leq \psi_n(x), n=1,2,3,\dots \quad \dots(1).$$

and

$$\int_E \psi_n(x)dx - \int_E \phi_n(x)dx < 1/n.$$

Since  $\phi_n$  are measurable,

$$\phi^* = \inf \phi_n \text{ is measurable. [by theorem 17.4]}$$

Similarly since  $\psi_n$  are measurable,

$$\psi^* = \sup \psi_n \text{ are measurable.}$$

Also from (1),

$$\begin{aligned} \text{Consider the set } \Delta_\nu &= \{x / \psi^*(x) < \phi^*(x)\} \\ &= \bigcup \Delta_\nu \end{aligned}$$

$$\text{where } \Delta_\nu = \{x / \phi^*(x) < \psi^*(x) - \frac{1}{\nu}\} \subset \{x / \phi_n(x) < \psi_n(x) - \frac{1}{\nu}\}.$$

From (1), the latter set has measure less than  $1/\nu$ .

Since  $\nu$  is arbitrary,  $m(\Delta_\nu) = 0$ .

Hence  $m(\Delta_\nu) = 0$ .

Hence  $\phi^* = \psi^*$  except on a set of measure zero.

Hence  $\phi^* = f$  except on a set of measure zero.

Hence  $f$  is measurable. [by theorem 17.5]

Hence the condition is also necessary.

Hence the theorem.

### 18.3 The Lebesgue integral of bounded functions over a set of finite measure

#### Definition

If  $f$  is a bounded measurable function defined on a measurable set  $E$  with  $mE$  finite, we define the Lebesgue integral of  $f$  over  $E$  by

$$\int_E f(x)dx = \inf \int_E \psi(x)dx \text{ for all simple functions } \psi \leq f.$$

Remark:

1.If  $f$  vanishes outside a set  $E$  of finite measure,

$$\text{we write } \int f \text{ for } \int_E f.$$

2. the following theorem shows that Lebesgue integral is the generalization of the Riemann integral.

#### Definition

A step function is a function which has the form

$$\psi(x) = c_i, \text{ if } x_{i-1} < x < x_i,$$

for some subdivision  $\{x_0, x_1, \dots, x_n\}$  of  $[a, b]$  and some constants  $c_i$ .

Therefore by the definition of integral,

$$\int_a^b \psi(x)dx = \sum_{i=1}^n c_i(x_i - x_{i-1}).$$

Hence upper Riemann integral of a function  $f$  can be defined as

$$R \int_a^b f(x)dx = \inf \int_a^b \psi(x)dx,$$

where the infimum is taken over all step functions  $\varphi(x) \leq f(x)$ .

And the lower Riemann integral of  $f$  is defined as

$$\underline{R} \int_a^b f(x) dx = \sup \int_a^b \varphi(x) dx ,$$

where the supremum is taken over all step functions  $\varphi(x) \leq f(x)$ .

Also we know that  $f$  is Riemann integrable if and only if

$$\overline{R} \int_a^b f(x) dx = \underline{R} \int_a^b f(x) dx$$

and the common value is denoted by  $R \int_a^b f(x) dx$ .

**Theorem 18.4**

Let  $f$  be a bounded function defined on  $[a, b]$ .

If  $f$  is Riemann integrable on  $[a, b]$ , then it is measurable and

$$R \int_a^b f(x) dx = \int_a^b f(x) dx .$$

**Proof**

Since every step function is a simple function,

$$\begin{aligned} \underline{R} \int_a^b f(x) dx &= \sup_{\varphi \leq f} \int_E \varphi(x) dx \\ &= \inf_{\psi \leq f} \int_E \psi(x) dx \\ &= R \int_a^b f(x) dx . \end{aligned}$$

Since  $f$  is Riemann integrable,

the inequalities are all equalities.

Hence  $f$  is measurable and

$$R \int_a^b f(x) dx = \int_a^b f(x) dx .$$

Hence the theorem.

**Theorem 18.5**

If  $f$  and  $g$  are bounded measurable functions defined on a set  $E$  of finite measure, then

- (i)  $\int_E (af + bg) = a \int_E f + b \int_E g$  .
- (ii) If  $f = g$  a.e., then  $\int_E f = \int_E g$  .
- (iii) If  $f \geq g$  a.e., then  $\int_E f \geq \int_E g$  .

$$\text{Hence } \left| \int_E f \right| \leq \int_E |f| .$$

- (iv) If  $A \subseteq E$  and  $B \subseteq E$ , then  $\int_A f + \int_B f = \int_{A \cup B} f$  .
- (v) If  $A$  and  $B$  are disjoint measurable sets of finite measure, then

$$\int_{A \cup B} f = \int_A f + \int_B f .$$

**Proof**

(i) To prove that  $\int_E (af + bg) = a \int_E f + b \int_E g$ .

If  $f$  is a simple function,

then  $af$  is also a simple function (if  $a \neq 0$ ).

Hence for  $a > 0$ ,

$$\begin{aligned} \int_E af &= \inf_{\psi \geq f} \int_E a\psi \\ &= a \inf_{\psi \geq f} \int_E \psi \\ &= a \int_E f. \end{aligned}$$

If  $a < 0$ , then

$$\begin{aligned} \int_E af &= \sup_{\phi \leq f} \int_E a\phi \\ &= a \inf_{\psi \geq f} \int_E \psi \\ &= a \int_E f. \end{aligned} \quad \text{[ by theorem 18.3]}$$

Hence if  $a \neq 0$ ,

$$\int_E af = a \int_E f. \quad \dots(1)$$

If  $f_1$  and  $f_2$  are simple functions such that  $f_1 \geq 0$  and  $f_2 \geq 0$ ,

then  $f_1 + f_2$  is also a simple function such that  $f_1 + f_2 \geq 0$ .

Hence

$$\begin{aligned} \int_E (f_1 + f_2) &= \inf \int_E (\psi_1 + \psi_2) \\ &= \inf \int_E (\psi_1 + \psi_2) \\ &= \int_E \psi_1 + \int_E \psi_2, \quad \text{for all } \psi_1 \geq f_1 \text{ and } \psi_2 \geq f_2. \end{aligned}$$

Hence,

$$\int_E (f_1 + f_2) = \inf \int_E \psi_1 + \inf \int_E \psi_2.$$

i.e  $\int_E (f_1 + f_2) = \int_E f_1 + \int_E f_2. \quad \dots(2)$

On the other hand,

if  $f_1$  and  $f_2$  are simple functions such that  $f_1 \leq 0$  and  $f_2 \leq 0$ ,

then  $f_1 + f_2$  is also a simple function such that  $f_1 + f_2 \leq 0$ .

Hence

$$\begin{aligned} \int_E (f_1 + f_2) &= \sup \int_E (\phi_1 + \phi_2) \\ &= \sup \int_E \phi_1 + \sup \int_E \phi_2, \quad \text{for all } \phi_1 \leq f_1 \text{ and } \phi_2 \leq f_2. \end{aligned}$$

Hence,

$$\int_E (f_1 + f_2) = \sup \int_E \phi_1 + \sup \int_E \phi_2.$$

Hence,

$$\int_E (f_1 + f_2) = \int_E f_1 + \int_E f_2 \quad \dots(3)$$

From (2) and (3), we get

$$\int_E (f_1 + f_2) = \int_E f_1 + \int_E f_2 \quad \dots(4)$$

Therefore from (1) and (4), we get

$$\begin{aligned} \int_E (af + bg) &= \int_E af + \int_E bg \\ &= a \int_E f + b \int_E g . \end{aligned}$$

Hence (i) is proved.

(ii) To prove that, if  $f = g$  a.e., then  $\int_E f = \int_E g$ .

Since  $f = g$  a.e.,  
 $f - g = 0$  a.e.

Hence,

if  $f - g \geq 0$ , then  $\int_E (f - g) = 0$  a.e.

Hence ,

$$\int_E \psi \leq 0, \quad \text{for all } \psi \geq f - g .$$

Therefore,

$$\inf \int_E \psi \leq 0 .$$

Hence,

$$\int_E (f - g) = 0 .$$

From (i),  $\int_E (f - g) = \int_E f - \int_E g$

Hence ,

$$\int_E f - \int_E g = 0 .$$

Hence ,

$$\int_E f = \int_E g \quad \dots(5).$$

Similarly,

if  $f - g \leq 0$ ,  
 then  $\int_E (f - g) = 0$  a.e

Hence ,

$$\int_E \varphi \geq 0 \quad \text{for all } \varphi \leq f - g .$$

Hence,

$$\sup \int_E \varphi \geq 0 .$$

Hence ,

$$\int_E (f - g) = 0 .$$

Hence,

$$\int_E f - \int_E g = 0 .$$

Hence ,

$$\int_E f = \int_E g \quad \dots(6)$$

From (5) and (6), we get

$$\int_E f = \int_E g .$$

Hence (ii) is proved.

(iii) To prove that, if  $f \geq g$  a.e., then  $\int_E f \geq \int_E g$ .

If  $f \geq g$  a.e.,  
then  $f - g \geq 0$  a.e.

Hence,

if  $f - g \geq 0$  a.e.,  
then  $\int_E (f - g) \geq 0$ .

Hence,

$$\int_E \varphi \geq 0 \quad \text{for all } \varphi = f - g.$$

Hence,

$$\sup \int_E \varphi \geq 0.$$

Hence,

$$\int_E (f - g) \geq 0.$$

Hence,

$$\int_E f - \int_E g \geq 0.$$

Hence,

$$\int_E f \geq \int_E g.$$

Since  $f \leq |f|$ ,

from this inequality we get,

$$\int_E f \leq \int_E |f|. \quad \dots(7)$$

Similarly since  $-f \leq |f|$ ,

$$-\int_E f \leq \int_E |f|. \quad \dots(8)$$

From (7) and (8),

$$|\int_E f| \leq \int_E |f|.$$

Hence (iii) is proved.

(iv) To prove that if  $A \leq f(x) \leq B$ , then  $A mE \leq \int_E f \leq B mE$ .

From (iii),

$$\text{if } A \leq f(x), \quad \int_E A \leq \int_E f. \quad \dots(9)$$

Since  $A$  is a constant,

$$\begin{aligned} \int_E A &= A \int_E 1 \\ &= A mE. \end{aligned}$$

Therefore from (9),

$$A mE \leq \int_E f. \quad \dots(10)$$

Similarly if  $f(x) \leq B$ ,

$$\begin{aligned} \text{then } \int_E f &\leq \int_E B \\ &= B mE \end{aligned} \quad \dots(11)$$

From (10) and (11), we get that

$$\text{if } A \leq f(x) \leq B, \quad \text{then } A mE \leq \int_E f \leq B mE.$$

Hence (iv) is proved.

(v) To prove that if  $A$  and  $B$  are disjoint measurable sets of finite measure, then

$$\int_{A \cup B} f = \int_A f + \int_B f.$$

Since  $A$  and  $B$  are measurable sets,

$A \cup B$  is also a measurable set.

Since  $A$  and  $B$  are also disjoint,

$$\chi_{A \cup B} = \chi_A + \chi_B.$$

Therefore

$$\begin{aligned} \int_{A \cup B} f &= \int f \cdot \chi_{A \cup B} \\ &= \int f \cdot (\chi_A + \chi_B) \\ &= \int f \cdot \chi_A + \int f \cdot \chi_B \\ &= \int_A f + \int_B f. \end{aligned}$$

Hence (v) is proved.

Hence the theorem.

### Theorem 18.6 Bounded convergence theorem

Let  $\{f_n\}$  be a sequence of measurable functions defined on a set  $E$  of finite measure, and suppose that there is a real number  $M$  such that

$$|f_n(x)| \leq M \text{ for all } n \text{ and all } x.$$

If  $f(x) = \lim_{n \rightarrow \infty} f_n(x)$  for each  $x$  in  $E$ , then

$$\int_E f = \lim_{n \rightarrow \infty} \int_E f_n.$$

#### Proof

Let  $\{f_n\}$  be a sequence of measurable functions defined on a set  $E$  of finite measure, and suppose that there is a real number  $M$  such that

$$|f_n(x)| \leq M, \text{ for all } n \text{ and all } x. \tag{1}$$

$$\text{Let } f(x) = \lim_{n \rightarrow \infty} f_n(x), \text{ for each } x \text{ in } E. \tag{2}$$

Hence by Littlewood's third principle,

$$\text{for given } \epsilon > 0 \text{ and } \delta = \epsilon/4M,$$

there is an integer  $N$  and a measurable set  $A \subset E$  such that

for  $n \geq N$  and  $x \in E - A$ ,

$$|f_n(x) - f(x)| < \epsilon/2m(E - A). \tag{3}$$

From (1), for all  $x$  in  $A$  and for all  $n$ ,

$$\begin{aligned} |f_n(x) - f(x)| &\leq |f_n(x)| + |f(x)| \\ &\leq M + M = 2M \end{aligned} \tag{4}$$

Hence

$$\begin{aligned} \left| \int_E f_n - \int_E f \right| &= \left| \int_E (f_n - f) \right| \\ &\leq \int_E |f_n - f| \\ &= \int_{E-A} |f_n - f| + \int_A |f_n - f| \\ &< \int_{E-A} (\epsilon/2m(E - A)) + \int_A (2M), \text{ for } n \geq N \quad \text{[From(3) and (4)]} \\ &= (\epsilon/2m(E - A))m(E - A) + (2M)mA \\ &< (\epsilon/2m(E - A))m(E - A) + (2M)(\epsilon/4M) \end{aligned}$$

$$= \frac{1}{2} + \frac{1}{2}$$

$$= 1, \quad \text{for } n \in \mathbb{N}.$$

Hence ,

$$\int_E f_n > \int_E f .$$

### Check your progress

Prove that a bounded function  $f$  on  $[a, b]$  is Riemann integrable if and only if the set of points at which  $f$  is discontinuous has measure zero.

### 18.4 Let us sum up

In this lesson we have studied

- The definition of a simple function
- The integral of a simple function
- Lebesgue integral of a function
- Properties of the integral and
- Bounded convergence theorem.

### 18.5 Lesson End Activities

1. Prove that the sum and product of two simple functions are simple.
2. Let  $f$  be a nonnegative measurable function. Show that  $\int f = 0 \Rightarrow f = 0$  a.e

### 18.6 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Wiley and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## Lesson 19 The integral of a non-negative function

### 19.0 Introduction

#### 19.1 Aims and Objectives

#### 19.2 The integral of a non-negative function

#### 19.3 Let us sum up.

#### 19.4 Lesson End Activities

#### 19.5 References

### 19.0 Introduction

In this lesson we are going to study about the definition and the properties of the integral of non-negative functions and some important theorems.

#### 19.1 Aims and Objectives

After studying this lesson, you would know

- How to find the integral of a non-negative function?
- Properties of the integral of non-negative functions.
- Fatou's Lemma
- Monotone convergence theorem and
- Definition of Integrable function over a measurable set .

#### 19.2 The integral of a non-negative function

##### Definition

If  $f$  is a non-negative measurable function defined on a measurable set  $E$ , the integral of  $f$  is defined as

$$\int_E f = \sup_{h \leq f} \int_E h,$$

where  $h$  is a bounded function such that

$$m\{x / h(x) > 0\} \text{ is finite.}$$

##### Theorem 19.1

If  $f$  and  $g$  are non-negative measurable functions, then:

- (i)  $\int_E cf = c \int_E f$ ,  $c > 0$ .
- (ii)  $\int_E f + g = \int_E f + \int_E g$ .
- (iii) If  $f \leq g$  a.e., then  $\int_E f \leq \int_E g$ .

##### Proof

- (i) To prove that  $\int_E cf = c \int_E f$ ,  $c > 0$ .

For  $c > 0$ , consider

$$\begin{aligned} \int_E cf &= \sup_{h \leq f} \int_E ch \\ &= c \sup_{h \leq f} \int_E h \\ &= c \int_E f. \end{aligned}$$

Hence (i) is proved.

(ii) To prove that  $\int_E f + g = \int_E f + \int_E g$ .

If  $h$  and  $k$  are bounded measurable functions such that

$$h(x) \leq f(x) \text{ and}$$

$$k(x) \leq g(x),$$

then  $h+k$  is also a bounded measurable function such that

$$h(x) + k(x) \leq f(x) + g(x).$$

Hence ,

$$\begin{aligned} \int_E h + \int_E k &= \int_E h + k \\ &\leq \sup \int_E h + k \\ &= \int_E f + g . \end{aligned}$$

Hence ,

$$\sup \int_E h + \sup \int_E k \leq \int_E f + g$$

Hence

$$\int_E f + \int_E g \leq \int_E f + g \quad \dots(1)$$

On the other hand,

let  $l$  be a bounded measurable function which vanishes outside a set of measure zero and which is not greater than  $f + g$ .

Define the functions  $h$  and  $k$  by setting

$$h(x) = \min(f(x), l(x))$$

and

$$k(x) = l(x) - h(x).$$

Then

$$h(x) \leq f(x) \text{ and}$$

$$k(x) \leq g(x).$$

Also  $h$  and  $k$  are bounded by the bound for  $l$  and vanish where  $l$  vanishes.

Hence ,

$$\int_E l = \int_E h + \int_E k$$

Hence,

$$\sup_{l \leq f+g} \int_E l \leq \int_E f + \int_E g .$$

Hence,

$$\int_E f + g \leq \int_E f + \int_E g \quad \dots(2)$$

From (1) and (2),

$$\int_E f + g = \int_E f + \int_E g .$$

Hence (ii) is proved.

(iii) To prove that if  $f \geq g$  a.e., then  $\int_E f \geq \int_E g$ .

If  $f \geq g$  a.e.,

then  $f - g \geq 0$  a.e.

Therefore if  $h$  is a bounded measurable function such that

$$h(x) \leq (f - g)(x),$$

then

$$h \geq 0 \text{ a.e.}$$

Hence,

$$\int_E h \geq 0.$$

Therefore,

$$\sup \int_E h = 0.$$

Hence ,

$$\int_E f - g = 0.$$

Hence ,

$$\int_E f - \int_E g = 0.$$

Hence,

$$\int_E f = \int_E g .$$

Hence (iii) is proved.

Hence the theorem.

**Theorem 19.2 Fatou’s Lemma:**

If  $\{ f_n \}$  is a sequence of nonnegative measurable functions and  $f_n(x) \geq f(x)$  almost everywhere on a set E, then

$$\int_E f \leq \underline{\lim} \int_E f_n .$$

**Proof**

Without loss of generality, assume that  $f_n(x) \geq f(x)$  everywhere, since integrals over a set of measure zero are zero.

Let h be a bounded measurable function which is not greater than f and which vanishes outside a set E’ of finite measure.

For  $n=1,2,3,\dots$

define a function  $h_n$  by setting

$$h_n(x) = \min \{ h(x), f_n(x) \}.$$

Then  $h_n$  is bounded by the bound for h and vanishes outside E’.

Also  $h_n(x) \geq h(x)$  for each x in E’.

Hence by Bounded convergence theorem,

$$\int_E h = \int_{E'} h = \lim \int_{E'} h_n = \underline{\lim} \int_E f_n .$$

This is true for all bounded measurable functions h which is not greater than f and which vanishes outside a set of finite measure.

Hence taking supremum over h,

$$\sup \int_E h = \underline{\lim} \int_E f_n .$$

Hence 
$$\int_E f \leq \underline{\lim} \int_E f_n .$$

Hence the theorem.

**Theorem 19.3 Monotone convergence theorem:**

Let  $\{f_n \}$  be an increasing sequence of nonnegative measurable functions, and let  $f = \lim f_n$  a.e.

Then 
$$\int f = \lim \int f_n .$$

**Proof**

By theorem 19.2,

$$\int f \leq \underline{\lim} \int f_n . \tag{1}$$

Since  $\{f_n\}$  is an increasing sequence nonnegative measurable functions, and  $f = \lim f_n$  a.e.,

$$\text{for each } n, f_n \leq f.$$

Hence 
$$\int f_n \leq \int f.$$

Therefore,

$$\overline{\lim} \int f_n \leq \int f \tag{2}$$

From (1) and (2),

$$\overline{\lim} \int f_n \leq \int f \leq \underline{\lim} \int f_n.$$

But we know that, it is always true that,

$$\underline{\lim} \int f_n \leq \overline{\lim} \int f_n \tag{3}$$

From (2) and (3),

$$\underline{\lim} \int f_n = \overline{\lim} \int f_n = \int f$$

Hence,

$$\int f = \lim \int f_n.$$

Hence the theorem.

**Theorem 19.4**

Let  $\{u_n\}$  be a sequence of nonnegative measurable functions, and

Let 
$$f = \sum_{n=1}^{\infty} u_n.$$

Then 
$$\int f = \sum_{n=1}^{\infty} \int u_n.$$

**Proof**

Let 
$$s_n = \sum_{k=1}^n u_k.$$

Then  $\{s_n\}$  is also a sequence of nonnegative measurable functions such that

$$s_n \geq f.$$

i.e  $f = \lim s_n.$

Hence by theorem 19.3,

$$\begin{aligned} \int f &= \lim \int s_n \\ &= \lim \int \sum_{k=1}^n u_k \\ &= \lim \sum_{k=1}^n \int u_k \\ &= \sum_{n=1}^{\infty} \int u_n. \end{aligned}$$

Hence,

$$\int f = \sum_{n=1}^{\infty} \int u_n.$$

Hence the theorem.

### Theorem 19.5

Let  $f$  be a nonnegative function and  $\{E_i\}$  be a disjoint sequence of measurable sets.

Let  $E = \cup E_i$ . Then

$$\int_E f = \sum \int_{E_i} f.$$

### Proof

Since  $E = \cup E_i$ ,

$$\chi_E = \sum \chi_{E_i}$$

Let  $u_i = f \cdot \chi_{E_i}$ .

$$\begin{aligned} \text{Then } f \cdot \chi_E &= f \cdot \sum \chi_{E_i} \\ &= \sum f \cdot \chi_{E_i} \\ &= \sum u_i. \end{aligned}$$

Hence by theorem 19.4,

$$\begin{aligned} \int_E f &= \int f \cdot \chi_E \\ &= \int \sum u_i \\ &= \sum \int u_i \\ &= \sum \int f \cdot \chi_{E_i} \\ &= \sum \int_{E_i} f. \end{aligned}$$

Hence the theorem.

### Definition

A nonnegative measurable function  $f$  is called integrable over the measurable set  $E$  if

$$\int_E f < \infty.$$

### Theorem 19.6

Let  $f$  and  $g$  be two nonnegative measurable functions.

If  $f$  is integrable over  $E$  and  $g(x) < f(x)$  on  $E$ ,

then  $g$  is also integrable on  $E$ , and

$$\int_E f - g = \int_E f - \int_E g.$$

### Proof

Since  $f = (f - g) + g$ , By theorem 19.1,

$$\int_E f = \int_E f - g + \int_E g.$$

Since the left side  $\int_E f$  is finite,

the terms on the right side is also finite.

Hence,

$$\int_E g \text{ is finite.}$$

Hence  $g$  is integrable and

$$\int_E f - g = \int_E f - \int_E g .$$

Hence the theorem.

**Theorem 19.7**

Let  $f$  be a nonnegative function which is integrable over a set  $E$ .  
 Then given  $\epsilon > 0$ ,  
 there is a  $\delta > 0$  such that for every set  $A \subset E$  with  $mA < \delta$ , we have

$$\int_A f < \epsilon .$$

**Proof**

If  $f$  is bounded, then the theorem is trivial.

Let  $f_n(x) = f(x)$  if  $f(x) \leq n$   
 and  $f_n(x) = n$  otherwise.

Then each  $f_n$  is bounded and  $f_n$  converges to  $f$  at each point.

Hence by monotone convergence theorem,

$$\int f = \lim \int f_n .$$

Hence there is an  $N$  such that

$$\int_E f_N > \int_E f - \epsilon / 2 .$$

Hence,

$$\begin{aligned} \int_E (f - f_N) &= \int_E f - \int_E f_N \\ &< \epsilon / 2 . \end{aligned}$$

Choose  $\delta < \epsilon / 2N$ .

Now if  $A \subset E$  with  $mA < \delta$ ,

$$\begin{aligned} \int_A f &= \int_A f - f_N + \int_A f_N \\ &< \int_E (f - f_N) + NmA \\ &< \epsilon / 2 + \epsilon / 2 = \epsilon . \end{aligned}$$

Hence given  $\epsilon > 0$  there is a  $\delta > 0$  such that for every set  $A \subset E$  with  $mA < \delta$ , we have

$$\int_A f < \epsilon .$$

Hence the theorem.

**Check your progress**

1. Show that we may have strict inequality in Fatou's Lemma.

Hint: consider the sequence  $\{f_n\}$  defined by  $f_n(x) = 1$  if  $n - x < n+1$ , with  $f_n(x) = 0$  otherwise.

2. Show that the monotone convergence theorem need not hold for decreasing sequence of functions.

Hint : Let  $f_n(x) = 0$ , if  $x < n$ ,  $f_n(x) = 1$  for  $x \geq n$ .

**19.3 Let us sum up**

In this lesson we have studied

- The definition of integral of nonnegative measurable function
- The definition of integrable function
- Properties of the integrals of nonnegative measurable function
- Fatou's lemma
- Monotone convergence theorem and

- a property of a nonnegative integrable function .

#### 19.4 Lesson End Activities

1. Show that if  $f$  and  $g$  are measurable &  $|f| \leq |g|$  a.e., and if  $g$  is integrable, then prove that  $f$  is integrable

#### 19.5 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Wiley and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

**LESSON – 20****THE GENERAL LEBESGUE INTEGRAL AND CONVERGENCE IN MEASURE.****Contents****20.0 Introduction****20.1 Aims and objectives****20.2 The General Lebesgue integral****20.3 Convergence in measure****20.4 Let us sum up****20.5 Lesson End Activities****20.6 References****20.0 Introduction**

In this lesson, you are going to study about the general Lebesgue integral, some of its properties, convergence in measure and theorems related to them.

**20.1 Aims and objectives**

After studying this lesson, you would know

- What is the positive part of a function?
- What is the negative part of a function?
- Definition of General Lebesgue integral of a measurable function
- Properties of Lebesgue integral .
- Lebesgue convergence theorem
- Generalization of Lebesgue convergence theorem
- Definition of convergence in measure of a sequence of measurable functions and
- Every sequence of measurable sequence that converges in measure contains a subsequence that converges almost everywhere.

**20.2 The General Lebesgue integral****Definition**

The positive part of a function  $f$  is  $f^+ = f \vee 0$

i.e  $f^+(x) = \max\{f(x), 0\}$

The negative part of a function is  $f^- = f \wedge 0$ .

i.e  $f^-(x) = \min\{f(x), 0\}$

Hence  $f = f^+ - f^-$ .

And  $|f| = f^+ + f^-$

**Definition**

A measurable function  $f$  is said to be integrable over  $E$  if  $f^+$  and  $f^-$  are both integrable over  $E$ .

Then the integral of  $f$  is defined as

$$\int_E f = \int_E f^+ - \int_E f^- .$$

**Theorem 20.1**

Let  $f$  and  $g$  are integrable over  $E$ . Then

- (i) The function  $cf$  is integrable over  $E$ , and

$$\int_E cf = c \int_E f .$$

(ii) The function  $f+g$  is integrable over  $E$ , and

$$\int_E f + g = \int_E f + \int_E g .$$

(iii) If  $f \geq g$  a.e., then  $\int_E f \geq \int_E g$ .

(iv) If  $A$  and  $B$  are disjoint measurable sets contained in  $E$ , then

$$\int_{A \cup B} f = \int_A f + \int_B f$$

**Proof**

(i) Since  $f$  is integrable over  $E$ , both  $f^+$  and  $f^-$  are integrable over  $E$  and the integral of  $f$  is given by

$$\int_E f = \int_E f^+ - \int_E f^- .$$

Hence,

both  $cf^+$  and  $cf^-$  are integrable over  $E$ , and

hence,

$cf = cf^+ - cf^-$  are integrable over  $E$  and

$$\begin{aligned} \int_E cf &= \int_E cf^+ - \int_E cf^- . \\ &= c \int_E f^+ - c \int_E f^- && \text{[by theorem 19.1]} \\ &= c \left[ \int_E f^+ - \int_E f^- \right] \\ &= c \int_E f . \end{aligned}$$

Hence (i) is proved.

(ii) Suppose if  $f_1$  and  $f_2$  are nonnegative integrable functions with  $f = f_1 - f_2$ ,

Then  $f^+ - f^- = f_1 - f_2$ .

Hence ,

$$f^+ + f_2 = f^- + f_1 .$$

Hence by theorem 19.1,

$$f^+ + \int f_2 = \int f^- + \int f_1 .$$

Therefore,

$$\begin{aligned} f &= f^+ - f^- \\ &= f_1 - f_2 . \end{aligned} \tag{1}$$

Since  $f$  and  $g$  are measurable,

$f^+, f^-, g^+, g^-$  are measurable.

Hence,

$f^+ + g^+, f^- + g^-$  are also measurable.

And  $f + g = (f^+ + g^+) - (f^- + g^-)$ .

Hence by (1) and theorem 19.1,

$$\begin{aligned} (f + g) &= (f^+ + g^+) - (f^- + g^-) \\ &= f^+ + g^+ - f^- - g^- \\ &= (f^+ - f^-) + (g^+ - g^-) \\ &= f + g . \end{aligned}$$

Hence (ii) is proved.

(iii) Since  $f \leq g$  a.e.,  
 $f^+ - f^- \leq g^+ - g^-$  a.e.,

hence ,

$$f^+ + g^- \leq g^+ + f^- \text{ a.e.,}$$

Hence by using the results of theorem 19.1

$$\int (f^+ + g^-) \leq \int (g^+ + f^-).$$

Hence

$$\int f^+ + \int g^- \leq \int g^+ + \int f^-.$$

Hence ,

$$\int f^+ - \int f^- \leq \int g^+ - \int g^-$$

Hence,

$$\int f \leq \int g.$$

Hence (iii) is proved.

(iv) Consider  $\int_{A \cup B} f = \int f \cdot \chi_{A \cup B}$   
 $= \int f \cdot (\chi_A + \chi_B)$   
 $= \int f \cdot \chi_A + \int f \cdot \chi_B$   
 $= \int_A f + \int_B f.$

Hence (iv) is proved.

Hence the theorem.

**Theorem 20.2 Lebesgue Convergence theorem.**

Let  $g$  be integrable over  $E$  and let  $\{ f_n \}$  be a sequence of measurable functions such that  $|f_n| \leq g$  on  $E$  and for almost all  $x$  in  $E$  we have  $f(x) = \lim f_n(x)$ . Then

$$\int_E f = \lim \int_E f_n.$$

**Proof**

Since  $|f_n| \leq g$  on  $E$ ,  
 $g - f_n$  is nonnegative and hence by Fatou's Lemma,

$$\int_E (g - f) \leq \underline{\lim} \int_E (g - f_n). \tag{1}$$

Since  $f(x) = \lim f_n(x)$  a.e. on  $E$  and  $|f_n| \leq g$  on  $E$ ,

$$|f| \leq g \text{ on } E.$$

Hence since  $g$  is integrable,

$f$  is also integrable.

Therefore,

$$\int_E (g - f) = \int_E g - \int_E f \tag{2}$$

Also,

$$\underline{\lim} \int_E (g - f_n) = \int_E g - \overline{\lim} \int_E f_n \tag{3}$$

Substituting (2) and (3) in (1), we get

$$\int_E g - \int_E f \leq \int_E g - \overline{\lim} \int_E f_n$$

Hence

$$\int_E f \leq \overline{\lim} \int_E f_n . \tag{4}$$

Similarly by considering  $g + f_n$ , we get

$$\int_E f \leq \underline{\lim} \int_E f_n \tag{5}$$

From (4) and (5), we get

$$\overline{\lim} \int_E f_n = \int_E f = \underline{\lim} \int_E f_n . \tag{6}$$

But it is always true that

$$\underline{\lim} \int_E f_n \leq \overline{\lim} \int_E f_n \tag{7}$$

From (6) and (7),

$$\int_E f = \lim \int_E f_n .$$

Hence the theorem.

Remark:

If we replace  $g$  by  $g_n$ 's, we get the following generalization of the Lebesgue Convergence theorem.

**Theorem 20.3**

Let  $\{g_n\}$  be a sequence of integrable functions which converges a.e to an integrable function  $g$ . Let  $\{f_n\}$  be a sequence of measurable functions such that

$$|f_n| \leq g_n \text{ and } \{g_n\} \text{ converges to } g \text{ a.e.}$$

If  $\int g = \lim \int g_n$ ,

then  $\int f = \lim \int f_n$ .

**Check your progress**

1. Show that if  $f$  is integrable over  $E$ , then so is  $|f|$  and

$$\left| \int_E f \right| \leq \int_E |f| .$$

Does the integrability of  $|f|$  imply that of  $f$ ?

2. Show that under the hypotheses of theorem 20.3, we have

$$\int |f_n - f| \rightarrow 0 .$$

3. Let  $\{f_n\}$  be a sequence of integrable functions such that  $f_n \rightarrow f$  a.e with  $f$  integrable. Then  $\int |f_n - f| \rightarrow 0$  if and only if  $\int |f_n| \rightarrow \int |f|$ .

**20.3 Convergence in measure**

**Definition**

A sequence  $\{f_n\}$  of measurable functions is said to converge to  $f$  in measure if, given  $\epsilon > 0$ , there is an  $N$  such that for all  $n \geq N$  we have

$$m\{x \in E / |f(x) - f_n(x)| \geq \epsilon\} < \epsilon .$$

Remark:

From this definition and littlewood's third principle, it is clear that,

If  $\{f_n\}$  is a sequence of measurable functions defined on a measurable set  $E$  of finite measure and  $f_n \rightarrow f$  a.e, then  $\{f_n\}$  converges to  $f$  in measure.

**Example**

Construct the sequence  $\{f_n\}$  as follows:

Let  $n=k+2, 0 \leq k < \infty$ , and

Set  $f_n(x) = 1$  if  $x \in [k2^{-k}, (k+1)2^{-k}]$

And  $f_n(x) = 0$  otherwise.

Then  $m\{x / |f_n(x)| > \epsilon\} = 2^{-k} \approx 2^{-n/2}$  [ since  $2^{-k} = 2^{-n/2}$  ]

Hence  $f_n \rightarrow 0$  in measure.

Remark:

Note that the sequence  $\{f_n(x)\}$  has the value 1 for arbitrarily large values of n.

Hence  $\{f_n(x)\}$  does not converge for any  $x$  in  $[0,1]$ .

**Theorem 20.4**

Let  $\{f_n\}$  be a sequence of measurable functions that converges in measure to f.

Then there is a subsequence  $\{f_{n_k}\}$  that converges to f almost everywhere.

**Proof**

Since  $\{f_n\}$  is a sequence of measurable functions that converges in measure to f,

Given  $\epsilon > 0$ ,

there is an integer n such that for all  $n \geq n_0$ ,

$$m\{x / |f(x) - f_n(x)| > \epsilon\} < \epsilon. \tag{1}$$

Let  $E_k = \{x / |f_{n_k}(x) - f(x)| > \epsilon\}$ .

Therefore,

$$\text{if } x \notin \bigcup_{v=k}^{\infty} E_v,$$

then  $|f_{n_v}(x) - f(x)| < \epsilon$  for  $v \geq k$ .

Therefore,

$$f_{n_v}(x) \rightarrow f(x).$$

Hence  $f_{n_v}(x) \rightarrow f(x)$  for any  $x \notin A = \bigcap_{k=1}^{\infty} \bigcup_{v=k}^{\infty} E_v$ .

But  $mA = m\left[\bigcup_{v=k}^{\infty} E_v\right]$

$$\begin{aligned} &= \sum_{v=k}^{\infty} mE_v \\ &= 2^{-k+1}. \end{aligned}$$

Hence  $mA = 0$

Hence there there is a subsequence  $\{f_{n_k}\}$  that converges to f almost everywhere.

**Theorem 20.5**

Let  $\{f_n\}$  be a sequence of measurable functions defined on a measurable set E of finite measure.

Then  $\{f_n\}$  converges to f in measure if and only if every subsequence of  $\{f_n\}$  has in turn a subsequence that converges almost everywhere to f.

**Theorem 20.6**

Fatou's lemma and the monotone and Lebesgue Convergence theorem remain valid if 'convergence a.e.' is replaced by 'convergence in measure'.

**Check your progress.**

1. Prove theorem 20.5
2. Prove theorem 20.6

**20.4 Let us sum up**

In this lesson , we have studied

- Definition of General Lebesgue integral of a measurable function
- Properties of Lebesgue integral .
- Lebesgue convergence theorem
- Generalization of Lebesgue convergence theorem
- Definition of convergence in measure of a sequence of measurable functions and
- Every sequence of measurable sequence that converges in measure contains a subsequence that converges almost everywhere.

**20.5 Lesson End Activities**

1. Show that if  $f$  is integrable over  $E$ , then  $|f|$  is also integrable over  $E$ . further

$$\left| \int_E f \right| \leq \int_E |f|$$

is the converse true?

**20.6 References**

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## Unit 5 Differentiation and Integration

### Lesson 21 Differentiation of monotone functions

#### 21.0 Introduction

#### 21.1 Aims and Objectives

#### 21.2 Differentiation of monotone functions

#### 21.3 Let us sum up

#### 21.4 Lesson End Activities

#### 21.5 References

#### 21.0 Introduction

In this lesson, we are going to study about Vitali covering of a set and the differentiation of monotone functions.

#### 21.1 Aims and Objectives

After studying this lesson, you would know,

- The Vitali covering of a set
- A monotone function on an interval is differentiable and
- The derivative of the monotone function is measurable.

#### 21.2 Differentiation of monotone functions

##### Definition: Vitali covering

Let  $J$  be a collection of intervals. We say that  $J$  covers a set  $E$  in the sense of Vitali, if for each  $\epsilon > 0$  and any  $x$  in  $E$ , there is an interval  $I \in J$

such that  $x \in I$  and  $l(I) < \epsilon$ .

The intervals may be open, closed or half open, but degenerate intervals consisting of only one point is not allowed.

##### Theorem 21.1 Vitali Lemma

Let  $E$  be a set of finite outer measure and  $J$  a collection of intervals that cover in the sense of Vitali. Then, given  $\epsilon > 0$ , there is a finite disjoint collection  $\{I_1, I_2, \dots, I_N\}$  of intervals in  $J$  such that

$$m^* \left[ E - \bigcup_{n=1}^N I_n \right] < \epsilon.$$

##### Proof

Without loss of generality, assume that the intervals in  $J$  are closed, for otherwise, we can replace each interval by its closure.

Since the measure of the set of end-points of the intervals  $\{I_1, I_2, \dots, I_N\}$  has measure zero, our assumption would not affect the result.

Let  $O$  be an open set of finite measure containing  $E$ .

Since  $J$  is a Vitali covering of  $E$  and  $E \subset O$ ,

we may assume that each  $I$  of  $J$  is contained in  $O$ .

Construct a sequence  $\{I_n\}$  of disjoint intervals of  $J$  by induction as follows.

1. Let  $I_1$  be any interval in  $J$ .
2. Suppose  $I_1, I_2, \dots, I_n$  are already been chosen.
3. Let  $k_n$  be the supremum of the lengths of the intervals of  $J$  that do not meet any of the intervals  $I_1, I_2, \dots, I_n$ .
4. Since each  $I_n \subset O$ ,  $k_n \rightarrow 0$  as  $n \rightarrow \infty$ .

5. Unless  $E \subset \bigcup_{i=1}^n I_i$ , we can find  $I_{n+1}$  in  $J$  with  $l(I_{n+1}) > (\frac{1}{2})k_n$  and

$I_{n+1}$  disjoint from  $I_1, I_2, \dots, I_n$ .

Hence we have a sequence  $\{I_n\}$  of disjoint intervals of  $J$ .

Since each  $I_n \subset O$ ,  
 $\bigcup I_n \subset O$ .

Hence,

$$\sum l(I_n) \leq mO < \infty.$$

Hence we can find an integer  $N$  such that

$$\sum_{n=N+1}^{\infty} l(I_n) < \frac{\epsilon}{5}.$$

Let  $R = E - \bigcup_{n=1}^N I_n$ .

Claim:  $m^*R < \epsilon$ .

Let  $x$  be any arbitrary point of  $R$ .

Then  $x \notin$  the closed set  $\bigcup_{n=1}^N I_n$ .

Hence we can find an interval  $I$  in  $J$  which contains  $x$  and whose length is so small that  $I$  does not meet any of the intervals  $I_1, I_2, \dots, I_N$ .

If  $I \cap I_i = \emptyset$  for  $i = 1, \dots, n$ ,

Then we must have,

$$l(I) < 2l(I_{n+1}).$$

Since  $\lim l(I_n) = 0$ ,

the interval  $I$  must meet atleast one of the intervals  $I_n$ .

Let  $n$  be the smallest integer such that  $I$  meets  $I_n$ .

Then  $n > N$  and  $l(I) < 2l(I_n)$ .

Since  $x$  is in  $I$  and  $I$  has a point in common with  $I_n$ ,

it follows that the distance from  $x$  to the midpoint of  $I_n$

$$\begin{aligned} \text{is at most } & l(I) + \frac{1}{2}l(I_n) < 2l(I_n) + \frac{1}{2}l(I_n) \\ & = \frac{5}{2}l(I_n). \end{aligned}$$

Let  $J_n$  be the interval having the same mid-point as  $I_n$  and five times its length.

Then  $x \in J_n$ .

Hence,

$$R \subset \bigcup_{n=N+1}^{\infty} J_n.$$

$$\begin{aligned} \text{And } m^*R & \leq \sum_{n=N+1}^{\infty} l(J_n) \\ & = 5 \sum_{n=N+1}^{\infty} l(I_n) \\ & < \epsilon. \end{aligned}$$

Hence given  $\epsilon > 0$ , there is a finite disjoint collection  $\{I_1, I_2, \dots, I_N\}$  of intervals in  $J$  such that

$$m^* \left[ E - \bigcup_{n=1}^N I_n \right] < \epsilon.$$

Hence the theorem.

### Definition

The derivatives of a function  $f$  at  $x$  is defined as follows:

$$D^+ f(x) = \overline{\lim}_{h \rightarrow 0^+} \frac{f(x+h) - f(x)}{h},$$

$$D^- f(x) = \overline{\lim}_{h \rightarrow 0^-} \frac{f(x) - f(x-h)}{h}$$

$$D_+ f(x) = \underline{\lim}_{h \rightarrow 0^+} \frac{f(x+h) - f(x)}{h}$$

$$D_- f(x) = \underline{\lim}_{h \rightarrow 0^-} \frac{f(x) - f(x+h)}{h}.$$

Remarks :

From the above definition, it is clear that

1.  $D^+ f(x) = D_+ f(x)$
2.  $D^- f(x) = D_- f(x)$ .

### Definition

If  $D^+ f(x) = D_+ f(x) = D^- f(x) = D_- f(x) = \pm$ , then  $f$  is said to be differentiable at  $x$  and  $f'(x)$  is defined as the common value of the derivatives at  $x$ .

If  $D^+ f(x) = D_+ f(x)$ , then  $f$  is said to have a right hand derivative at  $x$ , denoted by  $f'(x_+)$  and  $f'(x_+)$  is defined as their common value.

If  $D^- f(x) = D_- f(x)$ , then  $f$  is said to have a left-hand derivative at  $x$ , denoted by  $f'(x_-)$  and  $f'(x_-)$  is defined as their common value.

### Check your progress

Prove that if  $f$  is continuous on  $[a, b]$  and one of its derivatives (say  $D^+$ ) is everywhere nonnegative on  $(a, b)$ , then  $f$  is nondecreasing on  $[a, b]$ .

i.e  $f(x) \leq f(y)$  for  $x < y$ .

### Theorem 21.2

Let  $f$  be an increasing real-valued function on the interval  $[a, b]$ .

Then  $f$  is differentiable almost everywhere.

The derivative  $f'$  is measurable, and

$$\int_a^b f'(x) dx \leq f(b) - f(a).$$

### Proof

First let us show that the set of all points at which any of two derivatives are unequal has measure zero.

Let  $E = \{ x / D^+ f(x) > D^- f(x) \}$

Then  $E$  is the union of the sets  $E_{u,v}$  where

$$E_{u,v} = \{ x / D^+ f(x) > u > v > D^- f(x) \},$$

for all rationals  $u$  and  $v$ .

If we show that  $m^* E_{u,v} = 0$ ,  
 Then  $m^* E_{u,v} = 0$   
 and hence  $m^* E = 0$ .

Similarly we can prove for all other cases.

Claim:  $m^* E_{u,v} = 0$

Let  $s = m^* E_{u,v}$ .

Choosing  $\epsilon > 0$ , enclose  $E_{u,v}$  in an open set  $O$  with  $mO < s + \epsilon$ .

For each point  $x$  in  $E_{u,v}$ ,

there is an arbitrarily small interval  $[x-h, x]$  contained in  $O$  such that  
 $f(x) - f(x-h) < vh$ . [since  $D^- f(x) < v$ ]

By Vitali lemma,

We can choose a finite collection of intervals  $\{I_1, I_2, \dots, I_N\}$  of them  
 whose interiors cover a subset  $A$  of  $E_{u,v}$  of outer measure greater than  $s - \epsilon$ .  
 Summing over these intervals, we get

$$\begin{aligned} \sum_{n=1}^N [f(x_n) - f(x_n - h_n)] &< \sum_{n=1}^N v h_n \\ &= v \sum_{n=1}^N h_n \\ &< v mO \\ &< v(s + \epsilon). \end{aligned} \quad \dots(1)$$

Again for each point  $y$  in  $A$ ,

there is an arbitralily small interval  $(y, y+k)$  that is contained in some  $I_n$   
 and for which

$$f(y+k) - f(y) > uk, \quad [\text{since } D^+ f(x) > u]$$

Hence, again by vitali Lemma,

there is finite collection  $\{J_1, J_2, \dots, J_M\}$  of such intervals such that there union  
 contains a subset of  $A$  of outer measure greater than  $s - 2\epsilon$ .

Then summing over these intervals, we get

$$\sum_{i=1}^M [f(y_i + k_i) - f(y_i)] > \sum_{i=1}^M u k_i = u \sum_{i=1}^M k_i > u(s - 2\epsilon). \quad \dots(2)$$

Since each interval  $J_i$  is contained in some  $I_n$ ,

if we sum over those  $i$  for which  $J_i \subset I_n$ , we get

$$[f(y_i + k) - f(y)] \quad [f(x_n) - f(x_n - h_n)] \quad [\text{since } f \text{ is increasing}]$$

Hence From (1) and (2),

$$\begin{aligned} v(s + \epsilon) &> \sum_{n=1}^N [f(x_n) - f(x_n - h_n)] \\ &\quad \sum_{i=1}^M [f(y_i + k_i) - f(y_i)] \\ &> u(s - 2\epsilon) \end{aligned}$$

i.e  $v(s + \epsilon) > u(s - 2\epsilon)$ .

Since this is true for each  $\epsilon > 0$ , this implies

$$vs > us.$$

If  $s = 0$ , then we can cancel  $s$  on both sides and get

$$v > u,$$

which is a contradiction.



### **21.4 Lesson End Activities**

1. Show that  $D^+(f+g) \leq D^+(f) + D^+(g)$

### **21.5 References**

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## LESSON – 22

### DIFFERENTIATION OF AN INTEGRAL AND ABSOLUTE CONTINUITY

#### Contents

- 22.0 Introduction
- 22.1 Aims and Objectives
- 22.2 Functions of bounded variations
- 22.3 Differentiation of an integral
- 22.4 Absolute continuity
- 22.5 Let us sum up
- 22.6 Lesson End Activities
- 22.7 References

#### 22.0 Introduction

In this lesson we are going to study about functions of bounded variation , differentiation of an integral and absolute continuity of a function

#### 22.1 Aims and Objectives

After studying this lesson, you would know

- The definition of function of bounded variation
- The definition of indefinite integral
- Differential of an integral
- Absolute continuity and
- Every absolutely continuous function is the indefinite integral of its derivative.

#### 22.2 Functions of bounded variations

##### Definition: Function of bounded variation

Let  $f$  be a real valued function on the interval  $[a, b]$ , and let  $a = x_0 < x_1 < \dots < x_k = b$  be any subdivision of  $[a, b]$ .

Define 
$$p = \sum_{i=1}^k [f(x_i) - f(x_{i-1})]^+$$

$$n = \sum_{i=1}^k [f(x_i) - f(x_{i-1})]^-$$

$$t = n + p = \sum_{i=1}^k |f(x_i) - f(x_{i-1})|$$

Here we use the following notation :

$$r^+ = r, \text{ if } r \geq 0, \text{ otherwise } r^+ = 0.$$

Similarly  $r^- = r, \text{ if } r \leq 0, \text{ otherwise } r^- = 0.$

And  $|r| = r^+ + r^-.$

Let  $P = \sup p,$

$$N = \sup n,$$

$$T = \sup t,$$

where we take the suprema over all possible subdivisions of  $[a, b]$ .

Clearly  $P = T = P+N.$

We call  $P$ , the positive variation of  $f$  over  $[a, b]$ ,  
 $N$ , the negative variation of  $f$  over  $[a, b]$  and  
 $T$ , the total variation of  $f$  over  $[a, b]$ .

Since  $T$  depends on the interval  $[a, b]$  and the function  $f$ ,  
 we sometimes write  $T$  as  $T_a^b$ , or  $T_a^b(f)$ .

If  $T < \infty$ ,  $f$  is said to be of bounded variation over  $[a, b]$ .  
 This can also be written as  $f \in BV$ .

Remarks:

Here we say two important results on bounded variation without proof

1. A function  $f$  is of bounded variation on  $[a, b]$  if and only if  $f$  is the difference of two monotone functions on  $[a, b]$ .
2. If  $f$  is of bounded variation on  $[a, b]$ , then  $f'(x)$  exists for almost all  $x$  in  $[a, b]$ .

### 22.3 Differentiation of an integral

#### Definition: Indefinite integral

If  $f$  is an integrable function on  $[a, b]$ , the indefinite integral of  $f$  is defined to be the function  $F$  defined on  $[a, b]$  as

$$F(x) = \int_a^x f(t) dt$$

#### Theorem 22.1

If  $f$  is integrable on  $[a, b]$ , then the function  $F$  defined by

$$F(x) = \int_a^x f(t) dt$$

is a continuous function of bounded variation on  $[a, b]$ .

#### Proof

By theorem 19.7,  $F$  is a continuous function on  $[a, b]$

To show that  $F$  is of bounded variation :

Let  $a = x_0 < x_1 < \dots < x_k = b$  be any subdivision of  $[a, b]$ .

$$\begin{aligned} \text{Then } T_a^b(F) &= \sum_{i=1}^k |F(x_i) - F(x_{i-1})| \\ &= \sum_{i=1}^k \left| \int_{x_{i-1}}^{x_i} f(t) dt \right| \\ &= \sum_{i=1}^k \int_{x_{i-1}}^{x_i} |f(t)| dt \\ &= \int_a^b |f(t)| dt . \end{aligned}$$

Hence ,

$$T_a^b(F) = \int_a^b |f(t)| dt < \infty .$$

Hence  $F$  is of bounded variation over  $[a, b]$ .

Hence the theorem.

#### Theorem 22.2

If  $f$  is integrable on  $[a, b]$  and

$$\int_a^x f(t) dt = 0 \text{ for all } x \in [a, b],$$

then  $f(t) = 0$  a.e in  $[a, b]$ .

**Proof**

Suppose  $f(x) > 0$  on a set  $E$  of positive measure.  
Then by theorem 16.15,  
there is a closed set  $F \subset E$  with  $mF > 0$ .

Let  $O = (a, b) - F$ .

Then either,

$$\int_a^b f \neq 0, \text{ or else}$$

$$0 = \int_a^b f = \int_F f + \int_O f.$$

Therefore,

$$\int_O f = - \int_F f = 0.$$

But since  $O$  is an open set, it is the disjoint union of a countable collection  $\{(a_n, b_n)\}$  of open intervals.  
Hence by theorem 19.4,

$$\int_O f = \sum \int_{a_n}^{b_n} f = 0.$$

Hence for some  $n$ , we have

$$\int_{a_n}^{b_n} f > 0.$$

Therefore,

$$\text{either } \int_a^{a_n} f > 0.$$

$$\text{Or } \int_a^{b_n} f > 0.$$

Hence in any case,

if  $f$  is positive on a set of positive measure, then for some  $x \in [a, b]$ ,

$$\int_a^x f(t) dt > 0.$$

Similarly we can prove for  $f$  negative on a set of positive measure.  
Hence we are getting a contradiction.

Hence  $f(t) = 0$  a.e on  $[a, b]$ .

**Theorem 22.3**

If  $f$  is bounded and measurable on  $[a, b]$  and

$$F(x) = \int_a^x f(t) dt + F(a),$$

then  $F'(x) = f(x)$  for almost all  $x$  in  $[a, b]$ .

**Proof**

By theorem 22.1,

$F$  is of bounded variation over  $[a, b]$ .

Hence  $F'(x)$  exists for almost all  $x$  in  $[a, b]$  and

$F$  is continuous on  $[a, b]$ .

Since  $f$  is bounded on  $[a, b]$ ,

$$|f| \leq K, \text{ for some real } K.$$

Define  $f_n(x) = \frac{F(x+h) - f(x)}{h}$  with  $h=1/n$ .

Hence ,

$$f_n(x) = \frac{1}{h} \int_x^{x+h} f(t) dt \quad \text{with } h=1/n.$$

Hence,

$$\begin{aligned} |f_n(x)| &= \left| \frac{1}{h} \int_x^{x+h} f(t) dt \right| \\ &\leq \frac{1}{h} \int_x^{x+h} |f(t)| dt \\ &\leq \frac{1}{h} \int_x^{x+h} K dt \\ &= Kh/h \\ &= K. \end{aligned}$$

Hence ,

$$|f_n| \leq K.$$

Since  $f_n(x) \rightarrow F'(x)$  a.e.,

by the bounded convergence theorem, we get for  $c \in [a, b]$ ,

$$\begin{aligned} \int_a^c F'(x) dx &= \lim \int_a^c f_n(x) dx \\ &= \lim_{h \rightarrow 0} \frac{1}{h} \int_a^c (F(x+h) - F(x)) dx \\ &= \lim_{h \rightarrow 0} \left[ \frac{1}{h} \int_a^c F(x+h) dx - \frac{1}{h} \int_a^c F(x) dx \right] \\ &= \lim_{h \rightarrow 0} \left[ \frac{1}{h} \int_{a+h}^{c+h} F(x) dx - \frac{1}{h} \int_a^c F(x) dx \right] \\ &= \lim_{h \rightarrow 0} \left[ \frac{1}{h} \int_c^{c+h} F(x) dx - \frac{1}{h} \int_a^{a+h} F(x) dx \right] \\ &= F(c) - F(a) \\ &= \int_a^c f(x) dx . \quad \text{[Since F is continuous]} \end{aligned}$$

Hence,

$$\int_a^c \{F'(x) - f(x)\} dx = 0 \text{ for all } c \in [a, b].$$

Hence by theorem 22.2

$$F'(x) - f(x) = 0 \text{ a.e on } [a, b].$$

Hence

$$F'(x) = f(x) \text{ a.e.}$$

Hence the theorem.

**Theorem 22.4**

Let  $f$  be an integrable function on  $[a, b]$ , and suppose that

$$F(x) = F(a) + \int_a^x f(t) dt .$$

Then  $F'(x) = f(x)$  for almost all  $x$  in  $[a, b]$ .

**Proof**

Without loss of generality assume that  $f \geq 0$ .

Let  $f_n$  be defined as

$$\begin{aligned} f_n(x) &= f(x) & \text{if } f(x) \leq n, \\ \text{and } f_n(x) &= n & \text{if } f(x) > n. \end{aligned}$$

Hence,

$$f - f_n \geq 0.$$

Hence ,

$$G_n(x) = \int_a^x (f - f_n) \tag{1}$$

is an increasing function of x.

Hence it has a derivative almost everywhere and this derivative is nonnegative. [by theorem 21.2]

Applying theorem 22.3 to the bounded and measurable functions  $f_n$ , we get

$$\frac{d}{dx} \int_a^x f_n = f_n(x) \text{ a.e.,}$$

From(1),

$$G_n(x) = \int_a^x f - \int_a^x f_n .$$

Therefore,

$$\int_a^x f = G_n(x) + \int_a^x f_n .$$

i.e  $F(x) - F(a) = G_n(x) + \int_a^x f_n .$

Hence,

$$\begin{aligned} F'(x) &= \frac{d}{dx} G_n(x) + \frac{d}{dx} \int_a^x f_n \\ &= \frac{d}{dx} G_n(x) + f_n(x) \text{ a.e.} \\ &= f_n(x) \text{ a.e.} \end{aligned}$$

Since n is arbitrary,

$$F'(x) = f(x) \text{ a.e.}$$

Hence,

$$\int_a^b F'(x) dx = \int_a^b f(x) dx = F(b) - F(a). \tag{2}$$

But by theorem 21.2 ,

$$\int_a^b F'(x) dx = F(b) - F(a). \tag{3}$$

From (2) and (3), we get

$$\int_a^b F'(x) dx = F(b) - F(a) = \int_a^b f(x) dx$$

Hence ,

$$\int_a^b (F'(x) - f(x)) dx = 0$$

Since  $F'(x) - f(x) \geq 0$ , this implies that

$$F'(x) - f(x) = 0 \text{ a.e.}$$

Hence,

$$F'(x) = f(x) \text{ a.e.}$$

Hence the theorem.

### 22.4 Absolute continuity

#### Definition

A real valued function f defined on [a, b] is said to be absolutely continuous on [a, b] if, given  $\epsilon > 0$ , there is a  $\delta > 0$  such that

$$\sum_{i=1}^n |f(x_i') - f(x_i)| < \epsilon$$

for every finite collection  $\{(x_i, x_i')\}$  of nonoverlapping intervals with

$$\sum_{i=1}^n |x_i' - x_i| < \delta$$

**Theorem 22.5**

If  $f$  is absolutely continuous on  $[a, b]$ , then it is of bounded variation on  $[a, b]$ .

**Proof**

Since  $f$  is absolutely continuous on  $[a, b]$ , given  $\epsilon = 1$ , there exists a  $\delta > 0$  such that

$$\sum_{i=1}^n |f(x_i') - f(x_i)| < 1,$$

for every finite collection  $\{(x_i, x_i')\}$  of nonoverlapping intervals with

$$\sum_{i=1}^n |x_i' - x_i| < \delta$$

Let  $K$  be the largest integer less than  $1 + (b - a)/\delta$ .

Now any subdivision of  $[a, b]$  can be split into  $K$  sets of intervals, each of total length less than  $\delta$ , by inserting fresh division points, if necessary.

Hence for any subdivision,

$$\begin{aligned} t &= \sum_{i=1}^k |f(x_i) - f(x_{i-1})| \\ &\leq \sum_{i=1}^K |f(x_i) - f(x_{i-1})| \\ &< 1 \\ &\leq K \cdot \delta \end{aligned}$$

i.e  $t < K \cdot \delta$

Hence,

$$\sup t < K \cdot \delta$$

Hence,

$$T < K \cdot \delta$$

Hence  $f$  is of bounded variation on  $[a, b]$ .

Hence the theorem.

**Check your progress**

1. An absolutely continuous function is continuous.
2. The sum and difference of two absolutely continuous functions is absolutely continuous.
3. Prove that the product of two absolutely continuous functions is absolutely continuous.
4. Prove that if  $f$  is absolutely continuous, then  $f$  has a derivative almost everywhere.

**Theorem 22.6**

If  $f$  is absolutely continuous on  $[a, b]$  and  $f'(x) = 0$  a.e., then  $f$  is constant.

**Proof**

To prove that  $f$  is constant,

it is enough to show that  $f(a) = f(c)$  for any  $c \in [a, b]$ .

Let  $c \in [a, b]$ .

Let  $E \subset (a, c)$  be the set of measure  $c - a$  in which  $f'(x) = 0$ .

Let  $\epsilon$  and  $\delta$  be arbitrary positive numbers.

Since  $f$  is absolutely continuous,

given  $\epsilon > 0$ , there is a  $\delta > 0$  such that

$$\sum_{i=1}^n |f(x_i') - f(x_i)| < \epsilon \quad \dots(1)$$

for every finite collection  $\{(x_i, x_i')\}$  of nonoverlapping intervals with

$$\sum_{i=1}^n |x_i' - x_i| < \delta.$$

To each  $x$  in  $E$ ,

there is an arbitrarily small interval  $[x, x+h]$  contained in  $[a, c]$  such that

$$|f(x+h) - f(x)| < \delta. \quad \dots(2)$$

Then by Vitali Lemma,

we can find a finite collection  $\{[x_k, y_k]\}$  of nonoverlapping intervals of this kind which cover all of  $E$  except for a set of measure less than  $\epsilon$ .

If we label the  $x_k$  such that  $x_k < x_{k+1}$ , then

$$y_0 = a \quad x_1 < y_1 \quad x_2 < \dots < y_n \quad c = x_{n+1}$$

and

$$\sum_{k=0}^n |x_{k+1} - y_k| < \delta.$$

From (2),

$$\begin{aligned} \sum_{k=1}^n |f(y_k) - f(x_k)| &= \sum_{k=1}^n |f(y_k) - f(x_k)| \cdot (y_k - x_k) \\ &< \delta \cdot (c - a) \end{aligned} \quad \dots(3)$$

From (1),

$$\sum_{k=0}^n |f(x_{k+1}) - f(y_k)| < \epsilon \quad \dots(4)$$

Hence

$$\begin{aligned} |f(c) - f(a)| &= \left| \sum_{k=0}^n [f(x_{k+1}) - f(y_k)] + \sum_{k=1}^n [f(y_k) - f(x_k)] \right| \\ &\leq \sum_{k=0}^n |f(x_{k+1}) - f(y_k)| + \sum_{k=1}^n |f(y_k) - f(x_k)| \\ &< \epsilon + \delta \cdot (c - a). \end{aligned}$$

Since  $\epsilon$  and  $\delta$  are arbitrary positive numbers,

$$f(c) - f(a) = 0.$$

Hence  $f(c) = f(a)$  for any  $c \in [a, b]$ .

Hence  $f$  is constant on  $[a, b]$ .

Hence the theorem.

### Theorem 22.7

A function  $F$  is an indefinite integral if and only if it is absolutely continuous.

#### Proof

If  $F$  is an indefinite integral,

then by theorem 19.7,  $F$  is absolutely continuous.

Conversely,

assume that  $F$  is absolutely continuous on  $[a, b]$ .

Then  $F$  is of bounded variation on  $[a, b]$ .

Hence  $F$  can be written as a difference of two monotone increasing functions, say  $F_1$  and  $F_2$ .

i.e  $F(x) = F_1(x) - F_2(x)$ .

Since increasing functions are differentiable almost everywhere,

$F'(x) = F_1'(x) - F_2'(x)$  exists almost everywhere.

Hence,

$$|F'(x)| = |F_1'(x)| + |F_2'(x)|.$$

Hence  $\int |F'(x)| dx = \int |F_1'(x)| dx + \int |F_2'(x)| dx$   
 $F_1(b) + F_2(b) - F_1(a) - F_2(a)$  [by theorem 21.2]

Hence  $F'(x)$  is integrable.

Let  $G(x) = \int_a^x F'(t) dt$ .

Then  $G$  is absolutely continuous.

Hence,

$f = F - G$  is also absolutely continuous.

By theorem 22.4,

$$G'(x) = F'(x) \text{ a.e.}$$

Hence,

$$f'(x) = 0 \text{ a.e.}$$

Hence by theorem 22.6,

$f$  is constant and equal to  $F(a)$ .

Hence,

$$F(x) = G(x) + F(a)$$

Hence,

$$F(x) = \int_a^x F'(t) dt + F(a).$$

Hence  $F$  is an indefinite integral.

Hence the theorem.

### Theorem 22.8

Every absolutely continuous function is the indefinite integral of its derivative.

#### Proof

Let  $F$  be an absolutely continuous function on an interval  $[a, b]$

Then by theorem 22.7,

$F$  is an indefinite integral.

Hence there exists a integrable function  $f$  such that

$$F(x) = \int_a^x f(x) dx \quad a \leq x \leq b.$$

Then by theorem 22.4,

$$F'(x) = f(x).$$

Hence  $F$  is the indefinite integral of its derivative.

Hence the theorem.

### 22.5 Let us sum up

Thus in this lesson, we have seen

- The definition of function of bounded variation.
- The definition of indefinite integral.
- If a function  $f$  is integrable on  $[a, b]$ , then its indefinite integral is a continuous function of bounded variation on  $[a, b]$ .
- If  $f$  is bounded and measurable on  $[a, b]$ , then the derivative of its indefinite integral is itself almost everywhere.

- The same is true if  $f$  is an integrable function on  $[a, b]$ .
- The definition of Absolute continuity.
- If the derivative of an absolutely continuous function is zero almost everywhere, then it is constant.
- A function is an indefinite integral if and only if it is absolutely continuous and
- Every absolutely continuous function is the indefinite integral of its derivative.

### 22.6 Lesson End Activities

1. A monotone function  $f$  on  $[a,b]$  is called singular if  $f' = 0$  a.e. Show that any monotonic increasing function is the sum of an Absolutely continuous function and a singular function.

### 22.7 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## LESSON – 23

### THE MINKOWSKI AND HOLDER INEQUALITIES

#### Contents

- 23.0 Introduction**
- 23.1 Aims and Objectives**
- 23.2 The  $L^p$  Spaces**
- 23.3 Convex functions**
- 23.4 The Minkowski Inequality**
- 23.5 The Holder Inequality**
- 23.6 Let us sum up**
- 23.7 Lesson End Activities**
- 23.8 References**

#### 23.0 Introduction

In this lesson, we are going to study about  $L^p$  spaces, The Minkowski and Holder Inequalities.

#### 23.1 Aims and Objectives

After studying this lesson, you would know

- The definition of  $L^p$  spaces.
- The definition of norm in the  $L^p$  spaces.
- Convex functions
- The Minkowski Inequality and
- The Holder Inequality.

#### 23.2 The $L^p$ Spaces

##### Definition

Let  $p$  be a positive real number. A measurable function defined on  $[0, 1]$  is said to belong to the space  $L^p = L^p[0, 1]$

$$\text{if } \int_0^1 |f|^p < \infty.$$

Remarks:

1.  $L^1$  consists precisely of the Lebesgue integrable functions on  $[0, 1]$ .
2. Since  $|f + g|^p \leq 2^p(|f|^p + |g|^p)$ ,  
Sum of two functions in  $L^p$  is again in  $L^p$ .
3. If  $f \in L^p$ ,

$$\text{then } \int_0^1 |f|^p < \infty.$$

Therefore for any scalar  $\alpha$ ,

$$\int_0^1 |\alpha f|^p = |\alpha|^p \int_0^1 |f|^p < \infty.$$

Hence  $\alpha f \in L^p$ .

4. From 2 and 3, it follows that  $f + g \in L^p$ ,  
whenever  $f \in L^p$  and  $g \in L^p$ , and  $\alpha, \beta$ , any scalars.  
Hence  $L^p$  spaces are linear spaces.

**Definition**

For a function  $f \in L^p$ , the norm of  $f$  is defined as

$$\|f\| = \|f\|_p = \left\{ \int_0^1 |f|^p \right\}^{1/p}.$$

**Remarks**

1. In the coming sections, we are going to prove that  $\|f + g\| \leq \|f\| + \|g\|$ .
2. A linear space is said to be a normed linear space, if for every  $f$  in the space, a number  $\|f\|$  associated in such a way that
  - (i)  $\|f\| = 0$  if and only if  $f \equiv 0$
  - (ii)  $\|kf\| = |k| \|f\|$
  - (iii)  $\|f + g\| \leq \|f\| + \|g\|$ .

3. Norms on linear spaces do not satisfy property(i), because from  $\|f\| = 0$ , we can only conclude that  $f = 0$  a.e.

Hence we consider two measurable functions to be equivalent if they are equal almost everywhere.

If we do not distinguish between equivalent functions, then  $L^p$  spaces are normed linear spaces.

4.  $L^\infty$  is used to denote the space of all bounded measurable functions on  $[0, 1]$ .
5. The norm on  $L^\infty$  is defined as

$$\|f\| = \|f\|_\infty = \text{ess sup } |f(t)|,$$

where  $\text{ess sup } f(t)$  is the infimum of  $\sup g(t)$ , as the  $g$  ranges over all functions which are equal to  $f$  almost everywhere.

6. Hence  $\text{ess sup } f(t) = \inf \{ M / m\{t / f(t) > M\} = 0 \}$ .

**Check your progress**

1. Prove that  $\|f\| = 0$  if and only if  $f = 0$  a.e.
2. Prove that  $\|kf\| = |k| \|f\|$ .
3. Show that  $\|f + g\| \leq \|f\| + \|g\|$ .
4. Show that  $\|f + g\|_1 \leq \|f\|_1 + \|g\|_1$ .
5. Show that  $\int fg \leq \|f\|_1 \|g\|$ .

**23.3 Convex functions**

**Definition**

A function defined on an open interval  $(a, b)$  is said to be convex if for each  $x, y \in (a, b)$  and each  $\lambda, 0 \leq \lambda \leq 1$ ,

$$f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda) f(y).$$

**Check your progress**

Prove that  $f(t) = t^p$  is a convex function on  $[0, \infty)$  for  $1 \leq p < \infty$ .

**23.4 The Minkowski Inequality**

**Theorem 23.1 Minkowski Inequality for  $1 \leq p < \infty$ .**

If  $f$  and  $g$  are in  $L^p$  with  $1 \leq p < \infty$ , then so is  $f+g$  and

$$\|f + g\|_p \leq \|f\|_p + \|g\|_p.$$

If  $1 < p < \infty$ ,

then equality can hold if there are nonnegative constants  $\alpha$  and  $\beta$  such that

$$\alpha f = \beta g.$$

**Proof**

The result is trivial if  $p = \infty$ ,  $\|f\| = 0$  or  $\|g\| = 0$  as shown below:

Case (i)  $p = \infty$ .

Since  $\operatorname{ess\,sup} (f + g)(t) = \operatorname{ess\,sup} f(t) + \operatorname{ess\,sup} g(t)$ ,

Hence  $\|f + g\|_\infty = \|f\|_\infty + \|g\|_\infty$ .

Case (ii)  $\|f\| = 0$  or  $\|g\| = 0$

If  $\|f\| = 0$ ,  
then  $f = 0$  a.e.,

Hence,

$$f + g = g \text{ a.e.}$$

Hence,

$$\|f + g\|_p = \|g\|_p.$$

Similar result follows when  $\|g\| = 0$ .

Hence the inequality holds.

Now Assume that  $1 < p < \infty$  and  $\|f\| = 1$ ,  $\|g\| = 1$ .

Hence there are functions  $f_0$  and  $g_0$  such that

$$|f| = f_0 \text{ and } |g| = g_0,$$

$$\text{and } \|f_0\| = \|g_0\| = 1.$$

Let  $\lambda = \alpha / (\alpha + \beta)$ .

Then  $(1 - \lambda) = 1 - \alpha / (\alpha + \beta)$   
 $= \beta / (\alpha + \beta)$ .

Therefore,

$$\begin{aligned} |f(x) + g(x)|^p &= (|f(x)| + |g(x)|)^p \\ &= [f_0(x) + g_0(x)]^p \\ &= [\lambda f_0(x) + (1 - \lambda)g_0(x)]^p \\ &= (\alpha + \beta)^p [\lambda f_0(x) + (1 - \lambda)g_0(x)]^p \\ &= (\alpha + \beta)^p [\lambda f_0(x)^p + (1 - \lambda)g_0(x)^p] \end{aligned} \dots(1)$$

[ Since  $t^p$  is a convex function on  $[0, \infty)$  for  $1 < p < \infty$  ]

Integrating both sides of this inequality gives

$$\int_0^1 |f(x) + g(x)|^p dx \leq (\alpha + \beta)^p \lambda \int_0^1 f_0(x)^p dx + (\alpha + \beta)^p (1 - \lambda) \int_0^1 g_0(x)^p dx$$

Hence ,

$$\begin{aligned} \|f + g\|_p^p &= (\alpha + \beta)^p [ \|f_0\|_p^p + (1 - \lambda) \|g_0\|_p^p ] \\ &= (\alpha + \beta)^p [ \|f\|_p^p + \|g\|_p^p ] \end{aligned} \quad [ \text{since } \|f_0\| = \|g_0\| = 1. ]$$

Taking p-th root on both sides, we get

$$\|f + g\|_p = \|f\|_p + \|g\|_p.$$

Hence the required result.

If  $1 < p < \infty$ ,

the inequality (1) is strict unless  $f_0(x) = g_0(x)$  and  $\operatorname{sgn} f(x) = \operatorname{sgn} g(x)$ .

i.e  $f_0 = g_0$  a.e and  $\operatorname{sgn} f = \operatorname{sgn} g$  a.e.

i.e  $f = g$ .

Hence the theorem.

Remark:

The function  $\phi(t) = t^p$  is a concave function on  $[0, \infty)$ , for  $0 < p < 1$ .

Hence the proof above gives, mutalis mutandis, the following inequality

### Minkowski Inequality for $0 < p < 1$ .

Let  $f$  and  $g$  be two nonnegative functions which belong to the space  $L^p$  with  $0 < p < 1$ .

Then  $\|f + g\| \leq \|f\| + \|g\|$ .

### 23.5 The Holder Inequality

#### Theorem 23.2

Let  $1 < p < \infty$ . then for  $a, b, t$  nonnegative we have

$$(a + tb)^p \leq a^p + ptba^{p-1}.$$

**Proof**

Let  $\phi(t) = (a + tb)^p - a^p - ptba^{p-1}$ .

Then  $\phi(0) = 0$ ,

And  $\phi'(t) = p(a+tb)^{p-1}b - pba^{p-1}$ .

$$= pb[(a+tb)^{p-1} - a^{p-1}] \geq 0, \quad \text{for } p \geq 1 \text{ and } a, b, t \geq 0.$$

Hence,

$\phi(t)$  is increasing and hence nonnegative for  $t > 0$ .

Hence,

$$(a + tb)^p - a^p - ptba^{p-1} \leq 0 \quad \text{for } a, b, t \text{ nonnegative.}$$

Hence,

$$(a + tb)^p \leq a^p + ptba^{p-1} \quad \text{for } a, b, t \text{ nonnegative.}$$

Hence the theorem.

#### Theorem 23.3 Holder Inequality:

If  $p$  and  $q$  are nonnegative extended real numbers such that

$$\frac{1}{p} + \frac{1}{q} = 1,$$

and if  $f \in L^p$  and  $g \in L^q$ , then  $f \cdot g \in L^1$  and

$$\int |fg| \leq \|f\|_p \|g\|_q.$$

Equality holds if and only if for some constants  $\alpha$  and  $\beta$ , not both zero, we have  $|f|^p = \alpha |g|^q$  a.e.

**Proof**

First consider the case  $p=1$  and  $q=\infty$ .

Then  $\int |fg| \leq (\sup |g|) \int |f|$

$$\|g\|_\infty \|f\|_1.$$

Hence we get the required result.

Hence assume that  $1 < p < \infty$  and consequently  $1 < q < \infty$ .

It is sufficient to consider the case  $f \geq 0$  and  $g \geq 0$ .

Otherwise if necessary, we can replace  $f$  by  $|f|$  and  $g$  by  $|g|$ .

Let  $h(x) = g(x)^{q-1} = g(x)^{q/p}$ . [since  $q-1 = q/p$ .]

Hence,

$$\begin{aligned} g(x) &= h(x)^{p/q} \\ &= h(x)^{p-1}. \end{aligned}$$

Therefore,

$$\begin{aligned} \int f(x)g(x) &= \int f(x)h(x)^{p-1} \\ &= \int (h(x) + f(x))^p - h(x)^p \quad [\text{by the above theorem}] \end{aligned}$$

Hence ,

$$\begin{aligned} \frac{d}{dt} \int fg &= \int |h+tf|^p - \int h^p \\ &= \|h+tf\|^p - \|h\|^p \\ &= (\|h\| + t\|f\|)^p - \|h\|^p \quad [\text{by Minkowski inequality}] \end{aligned}$$

Differentiating both sides with respect to t , we get

$$p \int fg = p(\|h\| + t\|f\|)^{p-1} \|f\|.$$

Hence at t=0, we get

$$p \int fg = p\|h\|^{p-1} \|f\|.$$

Hence ,

$$\int fg = \|h\|^{p-1} \|f\| = \|g\| \|f\|.$$

Hence we get the required result,

$$\int |fg| = \|f\|_p \|g\|_q.$$

The Minkowski inequality used above is equal only if there exist nonnegative constants and such that

$$\begin{aligned} |h| &= |f| \text{ a.e.} \\ \text{i.e. } |g|^{q/p} &= |f| \text{ a.e.} \\ \text{i.e. } |f|^p &= |g|^q. \end{aligned}$$

Hence the theorem.

### 23.6 Let us sum up

In this lesson , we have studied

- $L^p$  spaces .
- Norm on  $L^p$  Spaces
- Convex function
- Minkowski inequality and
- Holder inequality.

### 23.7 Lesson End Activities

if  $f : (a,b) \rightarrow \mathbb{R}$  is a continuous function such that  $f((x+y)/2) \leq (f(x) + f(y)) / 2 \forall x,y \in (a,b)$ . Prove that f is convex.

### 23.8 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

**LESSON - 24****CONVERGENCE AND COMPLETENESS****Contents****24.0 Introduction****24.1 Aims and Objectives****24.2 Convergence and Completeness****24.3 Let us sum up****24.4 Lesson End Activities****24.5 References****24.0 Introduction**

In this lesson we are going to study about the convergence of a sequence of functions in a normed linear space, Cauchy sequence and Banach spaces.

**24.1 Aims and Objectives**

After studying this lesson, you would know

- Convergence of a sequence of functions in a normed linear space
- Pointwise convergence
- Cauchy sequence of functions
- Complete space
- Banach space
- Summable series and
- Riesz-Fischer Theorem.

**24.2 Convergence and Completeness****Definition**

A sequence  $\{f_n\}$  in a normed linear space is said to converge to an element  $f$  in the space if given  $\epsilon > 0$ , there is an  $N$  such that for all  $n > N$ ,

$$\|f - f_n\| < \epsilon.$$

If  $f_n$  converges to  $f$ ,

we write  $f = \lim f_n$  or  $f_n \rightarrow f$ .

Remark:

1. Another way of formulating the convergence of  $f_n$  to  $f$  is that:

$$f_n \rightarrow f \quad \text{if} \quad \|f_n - f\| \rightarrow 0.$$

2. Convergence in the space  $L^p$  is often referred to as: convergence in the mean of order  $p$ .

3. Hence a sequence of functions  $\{f_n\}$  is said to converge to  $f$  in the mean of order  $p$  if each  $f_n \in L^p$  and  $\|f - f_n\| \rightarrow 0$ .

4. Convergence in  $L$  is nearly uniform convergence.

**Definition**

A sequence of functions  $\{f_n\}$  is said to converge pointwise to  $f$  if for each  $x$ ,  

$$f(x) = \lim f_n(x).$$

**Definition**

A sequence of functions  $\{f_n\}$  is said to converge to  $f$  almost everywhere if there is a set of measure zero such that for each  $x$  in  $\tilde{E}$ ,

$$f(x) = \lim f_n(x).$$

**Definition**

A sequence  $\{f_n\}$  in a normed linear space is said to be a Cauchy sequence if given  $\epsilon > 0$ , there is an  $N$  such that for all  $n > N$  and all  $m > N$ ,

$$\|f_n - f_m\| < \epsilon.$$

**Definition**

A normed linear space is called complete if every Cauchy sequence in the space converges.

i.e, if for each Cauchy sequence  $\{f_n\}$  in the space there is an element  $f$  in the space such that  $f_n \rightarrow f$ .

**Definition**

A complete normed linear space is called a Banach space.

**Definition**

A series  $\{f_n\}$  in a normed linear space is said to be summable to a sum  $s$  if  $s$  is in the space and the sequence of partial sums of the series converges to  $s$ .

i.e.,  $\|s - \sum_{i=1}^n f_i\| \rightarrow 0$ .

We write this as  $s = \sum_{i=1}^{\infty} f_i$ .

**Definition**

The series  $\{f_n\}$  is said to be absolutely summable if  $\sum_{n=1}^{\infty} \|f_n\| < \infty$ .

Remark:

1. For a series of real numbers, absolute summability implies that the series is summable. But this is in general not true for series of elements in a normed linear space.
2. The following theorem shows that the implication holds when the space is complete.

**Theorem 24.1**

A normed linear space  $X$  is complete if and only if every absolutely summable series is summable.

**Proof**

Let  $X$  be a complete space.

To prove that every absolutely summable series is summable.

Let  $\{f_n\}$  be an absolutely summable series of elements of  $X$ .

i.e  $\sum_{n=1}^{\infty} \|f_n\| = M < \infty$ .

Hence for each  $\epsilon > 0$ , there is an  $N$  such that

$$\sum_{n=N}^{\infty} \|f_n\| < \epsilon.$$

Let  $s_n = \sum_{i=1}^n f_i$  be the partial sum of the series  $\{f_n\}$ .

Then for  $n, m \geq N$ ,

$$\begin{aligned} \|s_n - s_m\| &= \left\| \sum_{i=m}^n f_i \right\| \\ &\leq \sum_{i=m}^{\infty} \|f_i\| \\ &\leq \sum_{n=N}^{\infty} \|f_n\| \\ &< \epsilon. \end{aligned}$$

Hence the sequence  $\{s_n\}$  of partial sums is a Cauchy sequence in  $X$ .

Since  $X$  is complete,

$\{s_n\}$  converges to an element  $s$  in  $X$ .

Hence  $\{f_n\}$  is summable.

Conversely,

Assume that every absolutely summable series is summable.

To prove that  $X$  is complete.

Let  $\{f_n\}$  be a Cauchy sequence in  $X$ .

Then by definition,

for each integer  $k$ , there is an integer  $n_k$  such that

$$\|f_n - f_m\| < 2^{-k}, \quad \text{for all } n \geq n_k \text{ and } m \geq n_k.$$

The  $n_k$ 's may be chosen such that  $n_{k+1} > n_k$ .

Hence  $\{f_{n_k}\}_{k=1}^{\infty}$  is a subsequence of  $\{f_n\}$ .

Let  $g_1 = f_{n_1}$ , and

$$g_k = f_{n_k} - f_{n_{k-1}} \quad \text{for } k > 1.$$

Then,

$$\sum_{i=1}^k g_i = \sum_{i=1}^k (f_{n_i} - f_{n_{i-1}}) = f_{n_k}.$$

Hence,

$\{g_k\}$  is a series whose  $k$ -th partial sum is  $f_{n_k}$ .

But,

$$\|g_k\| = \|f_{n_k} - f_{n_{k-1}}\| < 2^{-k+1} \quad \text{if } k > 1.$$

Hence,

$$\begin{aligned} \|g_k\| &\leq \|g_1\| + 2^{-k+1} \\ &= \|g_1\| + 1. \end{aligned}$$

Hence the series  $\{g_k\}$  is absolutely summable.

Hence by assumption,

$\{g_k\}$  is also summable.

Hence there is an element  $f$  in  $X$  to which the partial sums of the series converge.

Hence the subsequence  $\{f_{n_k}\}$  converges to  $f$ .

Claim:  $f = \lim_{n \rightarrow \infty} f_n$ .

Since  $\{f_n\}$  is a Cauchy sequence,

for given  $\epsilon > 0$ , there is an  $N$  such that for all  $n \geq N$  and  $m \geq N$ ,

$$\|f_n - f_m\| < \epsilon/2.$$

Since  $f_{n_k} \rightarrow f$ ,

for the same  $\epsilon > 0$ , there is a  $K$  such that for all  $k \geq K$ ,

$$\|f_{n_k} - f\| < \epsilon/2.$$

Choose  $k$  large such that  $k > K$  and  $n_k \geq N$ .

Then for  $n > N$ ,

$$\begin{aligned} \|f_n - f\| &= \|f_n - f_{n_k} + f_{n_k} - f\| \\ &\leq \|f_n - f_{n_k}\| + \|f_{n_k} - f\| \\ &< \epsilon/2 + \epsilon/2 \\ &= \epsilon. \end{aligned}$$

Hence,

$$f_n \rightarrow f.$$

Hence every Cauchy sequence of points of  $X$  converges to an element in  $X$ .

Hence  $X$  is complete.

Hence the theorem.

**Theorem 24.2 Riesz-Fischer Theorem:**

The  $L^p$  spaces are complete.

**Proof**

Assume that  $1 < p < \infty$ .

By theorem 24.1, we need only to show that every absolutely summable series is summable in  $L^p$ .

Let  $\{f_n\}$  be an absolutely summable series in  $L^p$ .

$$\text{Let } \sum_{n=1}^{\infty} \|f_n\| = M < \infty.$$

Define the functions  $g_n$  as

$$g_n(x) = \sum_{k=1}^n |f_k(x)|.$$

By Minkowski inequality,

$$\begin{aligned} \|g_n\| &\leq \sum_{k=1}^n \|f_k\| \\ &\leq \sum_{n=1}^{\infty} \|f_n\| \\ &= M. \end{aligned}$$

i.e  $\|g_n\| \leq M$ .

Hence,

$$\int (g_n)^p \leq M^p.$$

For each  $x$ ,

$\{g_n(x)\}$  is an increasing sequence of extended real numbers

And hence,

$\{g_n(x)\}$  must converge to an extended real number.

Let it be  $g(x)$ .

The function  $g$  so defined is measurable and

since  $g_n \geq 0$ , by Fatou's Lemma

$$\int g^p \leq \liminf \int (g_n)^p \leq M^p.$$

Hence,

$g^p$  is integrable.

Hence  $g(x)$  is finite for almost all  $x$ .

For each  $x$  such that  $g(x)$  is finite,

the series  $\sum_{k=1}^{\infty} f_k(x)$  is an absolutely summable series of real numbers.

Hence it must be summable to a real number.

Define this number as  $s(x)$ .

Define  $s(x) = 0$  for those  $x$  where  $g(x) = \infty$ .

Hence,

the function  $s$  defined is the limit of the partial sums

$$s_n = \sum_{k=1}^n f_k \text{ almost everywhere.}$$

Hence,

$s$  is measurable.

Since  $|s_n(x)| \leq g(x)$ ,

$|s(x)| \leq g(x)$ .

Hence  $s$  is in  $L^p$  and

$$\begin{aligned} |s_n(x) - s(x)|^p &= (|s_n(x)| + |s(x)|)^p \\ &\leq (2g(x))^p \\ &= 2^p [g(x)]^p. \end{aligned}$$

Since  $2^p g^p$  is integrable and  $|s_n(x) - s(x)|^p$  converges to 0 for almost all  $x$ ,

$$\int |s_n - s|^p \rightarrow 0 \text{ [by Lebesgue convergence theorem]}$$

Hence,

$$\|s_n - s\|^p \rightarrow 0.$$

Hence,

$$\|s_n - s\| \rightarrow 0.$$

Hence the series  $\{f_n\}$  has the sum  $s$  in  $L^p$ .

Hence every absolutely summable series is summable in  $L^p$ .

Hence the  $L^p$  spaces are complete.

Hence the theorem.

### Check your progress

1. Prove that each convergent sequence is a Cauchy sequence.
2. Prove that  $L^p$  is complete.

### 24.3 Let us sum up

In this lesson, we have seen

- The definition of Convergence of a sequence of functions in a normed linear space
- The definition of Pointwise convergence
- The definition of Cauchy sequence of functions
- The definition of Complete space
- The definition of Banach space
- The definition of Summable series
- A space is complete if and only if every absolutely summable series is summable and
- The  $L^p$  spaces are complete.

### 24.4 Lesson End Activities

1. Prove that the space  $l_p$  is a Banach space given  $l_p$  ( $1 \leq p < \infty$ ) consist of all real sequences  $(a_n)$

$$\sum_{n=1}^{\infty} |a_n|^p < \infty \quad \& \quad ||\{a_n\}||_p = \left(\sum_{n=1}^{\infty} |a_n|^p\right)^{1/p}$$

### 24.5 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.

## LESSON - 25

### BOUNDED LINEAR FUNCTIONALS ON THE $L^p$ SPACES.

#### 25.0 Introduction

#### 25.1 Aims and Objectives

#### 25.2 Bounded linear functionals on the $L^p$ spaces.

#### 25.3 Let us sum up

#### 25.4 Lesson End Activities

#### 25.6 References

#### 25.0 Introduction

In this lesson, we are going to study about bounded linear functionals on the  $L^p$  spaces.

#### 25.1 Aims and Objectives

After studying this lesson, you would know

- Linear functionals
- Bounded linear functionals
- Bounded linear functionals on  $L^p$  spaces and
- Riesz Representation theorem.

#### 25.2 Bounded linear functionals on the $L^p$ spaces.

##### Definition

A linear functional on a normed linear space  $X$  is defined to be a mapping  $F$  of the space  $X$  into the set of real numbers such that

$$F(\alpha f + \beta g) = \alpha F(f) + \beta F(g)$$

where  $f, g \in X$  and  $\alpha, \beta$  are scalars.

##### Definition

A linear functional  $f$  is said to be bounded, if there is a constant  $M$  such that

$$|F(f)| \leq M \|f\|, \quad \text{for all } f \text{ in } X.$$

##### Definition

The smallest constant  $M$  for which the inequality

$$|F(f)| \leq M \|f\|, \quad \text{for all } f \text{ in } X,$$

is true is called the norm of  $F$ .

Hence,

$$\|F\| = \sup \left\{ \frac{|F(f)|}{\|f\|} \right\},$$

as  $f$  ranges over all nonzero elements of  $X$ .

#### Theorem 25.1

Each function  $g$  in  $L^q$  defines a bounded linear functional  $F$  on  $L^p$  by

$$F(f) = \int fg.$$

And  $\|F\| = \|g\|_q$ .

**Proof**

Let  $g$  be a function in  $L^q$ .

Define a functional  $F$  on  $L^p$  by

$$F(f) = \int fg, \quad \text{for all } f \text{ in } L^p.$$

Claim:  $F$  is a bounded linear functional on  $L^p$ .

For  $f_1, f_2 \in L^p$ , and for scalars  $\alpha, \beta$ , consider,

$$\begin{aligned} F(\alpha f_1 + \beta f_2) &= \int (\alpha f_1 + \beta f_2)g \\ &= \int \alpha f_1 g + \int \beta f_2 g \quad [\text{by the properties of integrals}] \\ &= \alpha F(f_1) + \beta F(f_2). \end{aligned}$$

Hence  $F$  is linear on  $L^p$ .

Consider  $|F(f)| = \left| \int fg \right|$

$$\leq \int |fg|$$

$$\leq \|f\|_p \|g\|_q \quad [\text{by Holder inequality}]$$

Hence,

$$\frac{|F(f)|}{\|f\|} \leq \|g\|_q.$$

Since this is true for all  $f$  in  $L^p$ ,

$$\|F\| = \sup \left\{ \frac{|F(f)|}{\|f\|} \right\} \leq \|g\|_q.$$

Hence,

$$F \text{ is a bounded linear functional with } \|F\| \leq \|g\|_q. \quad \dots(1)$$

To prove the opposite inequality,

Let  $f = |g|^{q/p} \text{sgn } g$ , for  $1 < p < \infty$ , where  $\text{sgn}$  is a function defined as

$$\begin{aligned} \text{sgn } x &= 1 \quad \text{if } x > 0, \\ \text{sgn } 0 &= 0, \text{ and} \\ \text{sgn } x &= -1 \quad \text{if } x < 0. \end{aligned}$$

Therefore,

$$\begin{aligned} |f|^p &= (|g|^{q/p} \text{sgn } g)^p \\ &= |g|^q \end{aligned}$$

And  $fg = (|g|^{q/p} \text{sgn } g)g$

$$= |g|^{q/p+1} \text{sgn } g$$

$$= |g|^q \quad [\text{since } 1/p + 1/q = 1]$$

Hence  $|f|^p = |g|^q = fg \quad \dots(2)$

Hence  $\int |f|^p = \int fg$ .

Hence  $f$  is in  $L^p$ .

And 
$$\begin{aligned} \|f\|_p &= \left(\int |f|^p\right)^{1/p} \\ &= \left(\int |g|^q\right)^{1/p} && \text{[from (2)]} \\ &= (\|g\|_q)^{q/p}. \end{aligned}$$

Therefore ,

$$\begin{aligned} F(f) &= \int fg \\ &= \int |g|^q && \text{[from (2)]} \\ &= (\|g\|_q)^q \\ &= \|g\|_q (\|g\|_q)^{q-1} \\ &= \|g\|_q (\|g\|_q)^{q/p} \\ &= \|g\|_q \|f\|_p. \end{aligned}$$

Hence ,

$$\frac{|F(f)|}{\|f\|_p} = \|g\|_q.$$

Hence,

$$\|F\| \text{ must be atleast as great as } \|g\|_q. \tag{3)}$$

From (1) and (3), we get

$$\|F\| = \|g\|_q.$$

Hence the theorem.

**Theorem 25.2**

Let  $g$  be an integrable function on  $[0, 1]$ , and suppose that there is a constant  $M$  such that

$$\left| \int fg \right| \leq M \|f\|_p$$

for all bounded measurable functions  $f$ .

Then  $g$  is in  $L^q$ , and  $\|g\|_q \leq M$ .

**Proof**

Assume that  $1 < p < \infty$ .

Define a sequence of bounded measurable functions  $g_n$  as

$$g_n(x) = \begin{cases} g(x) & \text{if } |g(x)| \leq n \\ 0 & \text{if } |g(x)| > n \end{cases}$$

Then  $\{g_n\}$  is a sequence of bounded measurable functions.

Let  $f_n = |g_n|^{q/p} \text{sgn } g_n$ .

Then 
$$\begin{aligned} |f_n|^p &= |g_n|^{q/p} \text{sgn } |g_n|^p \\ &= |g_n|^q. \end{aligned}$$

Therefore,

$$\begin{aligned} (\|f_n\|_p)^p &= \int |f_n|^p \\ &= \int |g_n|^q \\ &= (\|g_n\|_q)^q \end{aligned}$$

Hence,

$$\|f_n\|_p = (\|g_n\|_q)^{q/p} \tag{1)}$$

Also,

$$f_n g_n = (|g_n|^{q/p} \text{sgn } g_n) g_n$$

$$\begin{aligned} &= |g_n|^{q/p} |g_n| \\ &= |g_n|^{q/p+1} \\ &= |g_n|^q. \end{aligned}$$

Hence  $\int |g_n|^q = \int f_n g_n = \int f_n g$ . ...(2)

Therefore,

$$\begin{aligned} \int |g_n|^q &= \int f_n g \\ &= M \|f_n\|_p \|g_n\|_q \quad \text{[ by assumption]} \\ &= M (\|g_n\|_q)^{q/p} \quad \text{[ From (1)]} \end{aligned}$$

i.e  $(\|g_n\|_q)^q = M (\|g_n\|_q)^{q/p}$ .

Hence,

$$(\|g_n\|_q)^{q-q/p} = M.$$

Hence,

$$\|g_n\|_q = M^{1/p} \quad \text{[since } 1/p+1/q=1, \quad q-q/p=1]$$

Therefore,

$$\int |g_n|^q = M^q.$$

Since  $|g_n|^q$  converges to  $|g|^q$  a.e.,

By Fatou's Lemma,

$$\int |g|^q \leq \liminf \int |g_n|^q = M^q.$$

Hence ,

$$g \in L^q \text{ and } \|g\|_q = M.$$

Hence the theorem is proved for  $1 < p < \infty$ .

If  $p=1$ ,

$$\text{Let } E = \{ x \mid |g(x)| \leq M + \epsilon \}$$

$$\text{Let } f = (\text{sgn } g) \chi_E.$$

Then  $\|f\|_1 = mE$ .

Therefore,

$$\|fg\|_1 = \int |fg| = \int \chi_E |g| = mE.$$

Hence

$$\begin{aligned} M mE &= M \|f\|_1 \\ &= \int fg \quad \text{[ by assumption]} \\ &= (M + \epsilon) mE. \end{aligned}$$

Since  $\epsilon$  was arbitrary,

this implies  $mE = 0$ .

Hence ,

$$g \in L^1 \text{ and } \|g\|_1 = M$$

Hence we have proved the theorem for  $p=1$ .

Hence the theorem.

**Theorem 25.3 Riesz Representation theorem.**

Let  $F$  be a bounded linear functional on  $L^p, 1 \leq p < \infty$ .

Then there is a function  $g$  in  $L^q$  such that

$$F(f) = \int fg.$$

Also  $\|F\| = \|g\|_q$ .

**Proof**

Let  $\chi_s$  be the characteristic function of the interval  $[0, s]$ .  
 Since  $F$  is a bounded linear functional,  
 for each  $s$ ,  $F(\chi_s)$  is a real number.  
 Define  $\Phi(s)$  as this number.  
 Hence  $\Phi$  defines a function on  $[0, 1]$ .

Claim:  $\Phi$  is absolutely continuous.

Let  $\{(s_i, s_i')\}$  be any finite collection of nonoverlapping subintervals of  $[0, 1]$  of total length less than  $\epsilon$ .

Then  $\sum_i |\Phi(s_i') - \Phi(s_i)| = F(f)$ , where

$$f = \sum_i (\chi_{s_i'} - \chi_{s_i}) \operatorname{sgn}(\Phi(s_i') - \Phi(s_i)).$$

Hence 
$$(\|f\|_p)^p = \int |f|^p = \sum_i (s_i' - s_i) < \epsilon.$$

Hence 
$$\sum_i |\Phi(s_i') - \Phi(s_i)| = F(f) < \|F\| \|f\|_p < \|F\| \epsilon^{1/p}.$$

Therefore,

if we choose  $\epsilon = \epsilon^p / \|F\|^p$ , then

$$\sum_i |\Phi(s_i') - \Phi(s_i)| < \epsilon.$$

Hence  $\Phi$  is absolutely continuous.

Hence by theorem 22.7,

$\Phi$  is an indefinite integral.

i.e there is an integrable function  $g$  on  $[0, 1]$  such that

$$\Phi(s) = \int_0^s g$$

Hence 
$$\begin{aligned} F(\chi_s) &= \Phi(s) \\ &= \int_0^s g \\ &= \int_0^1 g \cdot \chi_s. \end{aligned}$$

Since every step function on  $[0,1]$  is equal almost everywhere to a suitable linear combination  $\sum c_i \chi_{s_i}$ ,

by the linearity of  $F$  and the integral, we get

$$F(\psi) = \int_0^1 g \psi$$

for each sep function  $\psi$ .

Let  $f$  be a bounded measurable function on  $[0, 1]$ .

By theorem 17.6,

there is a bounded sequence  $\{\psi_n\}$  of step functions which converge almost everywhere to  $f$ .

Hence the sequence  $\{|f - \psi_n|^p\}$  is uniformly bounded and

$$\|f - f_n\|^p > 0 \text{ a.e.}$$

Hence by bounded convergence theorem,

$$\|f - f_n\|^p > 0.$$

Since  $F$  is bounded, and

$$|F(f) - F(f_n)| = |F(f - f_n)| \leq \|F\| \|f - f_n\|_p.$$

Hence,

$$|F(f) - F(f_n)| > 0.$$

Hence,

$$F(f) = \lim F(f_n).$$

Since  $\|f - f_n\|_p$  is always less than  $\|g\|_q$  times the uniform bound for the sequence  $\{f_n\}$ , by bounded convergence theorem,

$$\int fg = \lim \int f_n g.$$

Hence,

$$\begin{aligned} \int fg &= \lim F(f_n) \\ &= F(f), \end{aligned}$$

for each bounded measurable function  $f$ .

Since  $|F(f)| \leq \|F\| \|f\|_p$ ,

$$|\int fg| \leq \|F\| \|f\|_p.$$

Hence by theorem 25.2,

$$g \in L^q \text{ and } \|g\|_q = \|F\|.$$

Claim:  $F(f) = \int fg$ , for each  $f$  in  $L^p$ .

Let  $f$  be an arbitrary function in  $L^p$ .

Then for each  $\epsilon > 0$ , there exists a step function  $f_n$  such that

$$\|f - f_n\|_p < \epsilon.$$

Since  $f_n$  is bounded,

$$F(f_n) = \int f_n g.$$

Hence

$$\begin{aligned} |F(f) - \int fg| &= |F(f) - F(f_n) + \int f_n g - \int fg| \\ &\leq |F(f) - F(f_n)| + |\int (f_n - f)g| \\ &\leq \|F\| \|f - f_n\|_p + \|g\|_q \|f - f_n\|_p \\ &< (\|F\| + \|g\|_q) \|f - f_n\|_p \\ &< (\|F\| + \|g\|_q) \epsilon. \end{aligned}$$

Since  $\epsilon$  is an arbitrary number,

$$F(f) = \int fg.$$

Hence by theorem 25.1,

$$\|F\| = \|g\|_q.$$

Hence, we have proved that there is a function  $g$  in  $L^q$  such that

$$F(f) = \int fg.$$

and  $\|F\| = \|g\|_q$ .

Hence the theorem.

### 25.3 Let us sum up

Thus in this lesson, we have seen,

- The definition of a linear functional on a normed linear space.
- Bounded linear functionals on  $L^p$
- For Each function  $g$  in  $L^q$ , there is a bounded linear functional  $F$  on  $L^p$  by
$$F(f) = \int fg \text{ such that } \|F\| = \|g\|_q.$$
- The converse of this is also true and it is given by Riesz Representation theorem.

#### 25.4 Lesson End Activities

1. Give the representation for the bounded linear functionals on  $l_p$  ( $1 \leq p < \infty$ )

#### 25.6 References

1. R.G. Bartle, Elements of Real Analysis, 2<sup>nd</sup> Edition, John Willy and Sons, New York, 1976.
2. W.Rudin, Real and Complex Analysis, 3<sup>rd</sup> Edition, McGraw-Hill, New York.