

# Лекция Т.2.4 — "2.4 Произведение мер"

## Введение в нелинейный функциональный анализ

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The Lebesgue measure on  $\mathbb{R}^N$  is in a sense the product of  $N$ -copies of the one-dimensional Lebesgue measure, since the volume of an  $N$ -dimensional rectangle is the product of the lengths of the sides. In this section we extend this idea to a general setting.

DEFINITION 2.4.1 Let  $(\Omega_1, \Sigma_1)$  and  $(\Omega_2, \Sigma_2)$  be two measurable spaces. For  $A \in \Sigma_1$  and  $B \in \Sigma_2$ , the set  $A \times B$  is said to be a "measurable rectangle".

The smallest  $\sigma$ -field of subsets of  $\Omega_1 \times \Omega_2$  which contains all measurable rectangles (i.e.  $\sigma(\mathcal{R})$  where  $\mathcal{R}$  is the collection of all measurable rectangles) is denoted by  $\Sigma_1 \times \Sigma_2$  and it is called the "product  $\sigma$ -field". Given  $C \subseteq \Omega_1 \times \Omega_2$  and  $\omega_1 \in \Omega_1$ ,  $\omega_2 \in \Omega_2$ , we set

$$C(\omega_1) = \{\omega_2 \in \Omega_2 : (\omega_1, \omega_2) \in C\}$$

and

$$C(\omega_2) = \{\omega_1 \in \Omega_1 : (\omega_1, \omega_2) \in C\}$$

and these sets are called the " $\Omega_1$ -section" and the " $\Omega_2$ -section" of  $C$ , respectively.

Similarly for a function  $f$  on  $\Omega_1 \times \Omega_2$  and  $\omega_1 \in \Omega_1$ ,  $\omega_2 \in \Omega_2$ , we put

$$f_{\omega_1}(\omega_2) = f(\omega_1, \omega_2) \quad \text{and} \quad f_{\omega_2}(\omega_1) = f(\omega_1, \omega_2)$$

and these functions of only one variable are called the " $\Omega_1$ -section" and the " $\Omega_2$ -section" of  $f$ , respectively.

REMARK 2.4.2 Note that  $\Sigma_1 \times \Sigma_2$  is not the Cartesian product of the  $\sigma$ -field, although the notation may suggest so.

If

$$C, C_i \subseteq \Omega_1 \times \Omega_2, \quad i \in I$$

and  $\omega_1 \in \Omega_1$ , then

$$\bigcup_{i \in I} C_i(\omega_1) = \left( \bigcup_{i \in I} C_i \right) (\omega_1),$$

$$\bigcap_{i \in I} C_i(\omega_1) = \left( \bigcap_{i \in I} C_i \right) (\omega_1),$$

$$(C^c)(\omega_1) = C(\omega_1)^c$$

and

$$(\chi_C)(\omega_1) = \chi_{C(\omega_1)}.$$

PROPOSITION 2.4.3 If  $(\Omega_1, \Sigma_1)$  and  $(\Omega_2, \Sigma_2)$  are measurable spaces,  $C \in \Sigma_1 \times \Sigma_2$  and  $f: \Omega_1 \times \Omega_2 \rightarrow \mathbb{R}^* = \mathbb{R} \cup \{\pm\infty\}$  is  $\Sigma_1 \times \Sigma_2$ -measurable function, then

(a) for every  $\omega_1 \in \Omega_1$  and every  $\omega_2 \in \Omega_2$  we have

$$C(\omega_1) \in \Sigma_2 \quad C(\omega_2) \in \Sigma_1;$$

(b) for every  $\omega_1 \in \Omega_1$ ,  $f_{\omega_1}$  is  $\Sigma_2$ -measurable  
and for every  $\omega_2 \in \Omega_2$ ,  $f_{\omega_2}$  is  $\Sigma_1$ -measurable.

Before stating and proving the first theorem of this section, we state a lemma whose proof is straightforward and it is left to the reader.

LEMMA 2.4.4 If  $(\Omega_1, \Sigma_1)$  and  $(\Omega_2, \Sigma_2)$  are measurable spaces, then

- (a) the family  $\mathcal{F}$  of all finite pairwise disjoint unions of measurable rectangles is a field of subsets of  $\Omega_1 \times \Omega_2$ ;
- (b)  $\Sigma_1 \times \Sigma_2$  is the smallest monotone class containing  $\mathcal{F}$ .

Примечание.

Families that satisfy (a) and (b) of Proposition 2.1.9 are called "monotone classes":

- (a) if  $\{A_n\}_{n \geq 1} \subseteq \mathcal{S}$  is increasing, then  $\bigcup_{n \geq 1} A_n \in \mathcal{S}$ ; or
- (b) if  $\{A_n\}_{n \geq 1} \subseteq \mathcal{S}$  is decreasing, then  $\bigcap_{n \geq 1} A_n \in \mathcal{S}$ .

THEOREM 2.4.5 If  $(\Omega_1, \Sigma_1, \mu_1)$  and  $(\Omega_2, \Sigma_2, \mu_2)$  are  $\sigma$ -finite measure spaces and  $C \in \Sigma_1 \times \Sigma_2$ , then

(a)  $\omega_1 \rightarrow \mu_2(C(\omega_1))$  is  $\Sigma_1$ -measurable;

(b)  $\omega_2 \rightarrow \mu_1(C(\omega_2))$  is  $\Sigma_2$ -measurable;

(c)

$$\int_{\Omega_1} \mu_2(C(\omega_1)) d\mu_1 = \int_{\Omega_2} \mu_1(C(\omega_2)) d\mu_2.$$



This theorem leads to the definition of the product measure which is made through the next result.

**THEOREM 2.4.6** If  $(\Omega_1, \Sigma_1, \mu_1)$  and  $(\Omega_2, \Sigma_2, \mu_2)$  are  $\sigma$ -finite measure spaces and

$$\mu_1 \times \mu_2: \Sigma_1 \times \Sigma_2 \rightarrow \overline{\mathbb{R}}_+ = \mathbb{R}_+ \cup \{+\infty\}$$

is defined by

$$(\mu_1 \times \mu_2)(C) = \int_{\Omega_1} \mu_2(C(\omega_1)) d\mu_1 = \int_{\Omega_2} \mu_1(C(\omega_2)) d\mu_2.$$

(see Theorem 2.4.5(c)),

then

$\mu_1 \times \mu_2$  is a  $\sigma$ -finite measure and for every measurable rectangle  $A \times B$ , we have

$$(\mu_1 \times \mu_2)(A \times B) = \mu_1(A)\mu_2(B)$$

(recall that  $0 \cdot \infty = 0$ ).

DEFINITION 2.4.7 The  $\sigma$ -finite measure

$$\mu_1 \times \mu_2: \Sigma_1 \times \Sigma_2 \rightarrow \overline{\mathbb{R}}_+ = \mathbb{R}_+ \cup \{+\infty\}$$

established in Theorem 2.4.6 is called the "product" of  $\mu_1$  and  $\mu_2$ .

For every measurable rectangle  $A \times B$  we have

$$(\mu_1 \times \mu_2)(A \times B) = \mu_1(A)\mu_2(B).$$

REMARK 2.4.8 The product measure  $\mu_1 \times \mu_2$  is uniquely determined by the requirements that it is a measure on  $\Sigma_1 \times \Sigma_2$  and that

$$(\mu_1 \times \mu_2)(A \times B) = \mu_1(A)\mu_2(B).$$

for every measurable rectangle  $A \times B$ .

Now we are ready to examine the relations between integrals on a product space and integrals on the component spaces.

The basic result in this direction is the "Fubini-Tonelli theorem", which enables us to evaluate integrals with respect to product measures in terms of iterated integrals.

PROPOSITION 2.4.9 If  $(\Omega_1, \Sigma_1, \mu_1)$  and  $(\Omega_2, \Sigma_2, \mu_2)$  are  $\sigma$ -finite measure spaces and

$$f: \Omega_1 \times \Omega_2 \rightarrow \overline{\mathbb{R}}_+ = \mathbb{R}_+ \cup \{+\infty\}$$

is  $\Sigma_1 \times \Sigma_2$ -measurable function, then

(a)

$$\omega_2 \mapsto \int_{\Omega_1} f(\omega_1, \omega_2) d\mu_1$$

is  $\Sigma_2$ -measurable and

$$\omega_1 \mapsto \int_{\Omega_2} f(\omega_1, \omega_2) d\mu_2$$

is  $\Sigma_1$ -measurable;

(b)

$$\int_{\Omega_1 \times \Omega_2} f d(\mu_1 \times \mu_2) = \int_{\Omega_2} \left( \int_{\Omega_1} f_{\omega_2}(\omega_1) d\mu_1 \right) d\mu_2 = \int_{\Omega_1} \left( \int_{\Omega_2} f_{\omega_1}(\omega_2) d\mu_2 \right) d\mu_1$$

## THEOREM 2.4.10 (Fubini-Tonelli Theorem)

If  $(\Omega_1, \Sigma_1, \mu_1)$  and  $(\Omega_2, \Sigma_2, \mu_2)$  are  $\sigma$ -finite measure spaces,

$$f \in \mathcal{L}(\Omega_1 \times \Omega_2, \mu_1 \times \mu_2)$$

and

$$h_1(\omega_1) = \int_{\Omega_2} f(\omega_1, \omega_2) d\mu_2, \quad h_2(\omega_2) = \int_{\Omega_1} f(\omega_1, \omega_2) d\mu_1,$$

then

$$h_1 \in \mathcal{L}(\Omega_1, \mu_1), \quad h_2 \in \mathcal{L}(\Omega_2, \mu_2)$$

and

$$\int_{\Omega_1 \times \Omega_2} f d(\mu_1 \times \mu_2) = \int_{\Omega_1} h_1 d\mu_1 = \int_{\Omega_2} h_2 d\mu_2.$$

REMARK 2.4.11 The measure space  $(\Omega_1 \times \Omega_2, \Sigma_1 \times \Sigma_2, \mu_1 \times \mu_2)$  is seldom complete, even if  $(\Omega_1, \Sigma_1, \mu_1)$  and  $(\Omega_2, \Sigma_2, \mu_2)$  are both complete measure spaces.

Indeed, if complete  $(\Omega_1, \Sigma_1, \mu_1)$ ,  $(\Omega_2, \Sigma_2, \mu_2)$  are two  $\sigma$ -finite measure spaces such that there exist

$$A \subseteq \Omega_1 \quad A \notin \Sigma_1$$

and a nonempty set  $B \in \Sigma_2$  with  $\mu_2(B) = 0$ , then the  $\sigma$ -finite measure space

$$(\Omega_1 \times \Omega_2, \Sigma_1 \times \Sigma_2, \mu_1 \times \mu_2)$$

is incomplete.

In particular, the  $\sigma$ -finite measure space

$$(\mathbb{R}^2, \mathcal{L} \times \mathcal{L}, \lambda_2 = \lambda \times \lambda)$$

is incomplete.

Let  $(\Omega_1 \times \Omega_2, \overline{\Sigma_1 \times \Sigma_2}, \overline{\mu_1 \times \mu_2})$  denote the completion of the measure space  $(\Omega_1 \times \Omega_2, \Sigma_1 \times \Sigma_2, \mu_1 \times \mu_2)$  (see Proposition 2.1.36).

The next lemma allows us to extend the Fubini-Tonelli theorem to this completed product measure space.

LEMMA 2.4.12 If  $(\Omega_1, \Sigma_1, \mu_1)$  and  $(\Omega_2, \Sigma_2, \mu_2)$  are  $\sigma$ -finite measure spaces,  $C \in \Sigma_1 \times \Sigma_2$  with

$$(\mu_1 \times \mu_2)(C) = 0$$

and  $D \subseteq C$ ,

then

$$\mu_1(D(\omega_2)) = 0$$

$\mu_2$ -a.e. on  $\Omega_2$  and

$$\mu_2(D(\omega_1)) = 0$$

$\mu_1$ -a.e. on  $\Omega_1$

## THEOREM 2.4.13 (Extended Fubini-Tonelli Theorem)

If  $(\Omega_1, \Sigma_1, \mu_1)$  and  $(\Omega_2, \Sigma_2, \mu_2)$  are complete  $\sigma$ -finite measure spaces and  $f: \Omega_1 \times \Omega_2 \rightarrow \overline{\mathbb{R}}_+ = \mathbb{R}_+ \cup \{+\infty\}$  is a  $\overline{\Sigma_1 \times \Sigma_2}$ -measurable function, then

(a) for  $\mu_2$ -almost all  $\omega_2 \in \Omega_2$  the function

$$\omega_1 \mapsto f(\omega_1, \omega_2)$$

is  $\Sigma_1$ -measurable and for  $\mu_1$ -almost all  $\omega_1 \in \Omega_1$  the function

$$\omega_2 \mapsto f(\omega_1, \omega_2)$$

is  $\Sigma_2$ -measurable;

(b) the function

$$\omega_2 \mapsto \int_{\Omega_1} f(\omega_1, \omega_2) d\mu_1$$

is  $\Sigma_2$ -measurable and the function

$$\omega_1 \mapsto \int_{\Omega_2} f(\omega_1, \omega_2) d\mu_2$$

is  $\Sigma_1$ -measurable; (см. продолжение на следующем слайде:)



## Теорема 2.4.13

Условие (с)

$$(c) \quad \int_{\Omega_1 \times \Omega_2} f d(\overline{\mu_1 \times \mu_2}) = \int_{\Omega_2} \left( \int_{\Omega_1} f_{\omega_2}(\omega_1) d\mu_1 \right) d\mu_2 = \\ \int_{\Omega_1} \left( \int_{\Omega_2} f_{\omega_1}(\omega_2) d\mu_2 \right) d\mu_1.$$

Next we will present a generalized version of the Fubini-Tonelli theorem using transition measures.

Transition measures (in particular transition probabilities) are the main tool in the relaxation of control systems (Section A.4.1) and in stochastic games (Section A.5.5).

DEFINITION 2.4.14 Let  $(\Omega_1, \Sigma_1)$  and  $(\Omega_2, \Sigma_2)$  be two measurable spaces. A map  $m: \Omega_1 \times \Sigma_2 \rightarrow \overline{\mathbb{R}}_+ = \mathbb{R}_+ \cup \{+\infty\}$  is said to be a "transition measure" if

(a) for every  $B \in \Sigma_2$ , the function  $\omega_1 \mapsto m(\omega_1, B)$  is  $\Sigma_1$ -measurable;

(b) for every  $\omega_1 \in \Omega_1$ , the set function  $B \mapsto m(\omega_1, B)$  is  $\sigma$ -finite measure on  $\Sigma_2$ .

We say that  $m$  is a " $\sigma$ -finite transition measure" if

$$\Omega_2 = \bigcup_{n \geq 1} A_n, \quad A_n \in \Sigma_2$$

and for all  $\omega_1 \in \Omega_1$  we have

$$m(\omega_1, A_n) < +\infty$$

for all  $n \geq 1$ .

Finally we say that  $m$  is a "transition probability", if for all  $\omega_1 \in \Omega_1$  we have

$$m(\omega_1, \Omega_2) = 1.$$

REMARK 2.4.15 Transition probabilities have the following interpretation: there are two systems whose states are described by the points in the sets  $\Omega_1$  and  $\Omega_2$ .

The statistical behavior of the outcomes of the second system depends on the state of the first system.

If the state of the first system is  $\omega_1 \in \Omega_1$ , then the probability that the state of the second system is in a set  $B$  is described by the transition probability  $m(\omega_1, B)$ .

EXAMPLES 2.4.16 (a) Let  $\mu$  be a  $\sigma$ -finite measure on  $(\Omega_2, \Sigma_2)$  and set

$$m(\omega_1, B) = \mu(B)$$

for all  $\omega_1 \in \Omega_1$  and all  $B \in \Sigma_2$ . Then  $m$  is trivially a transition measure.

(b) Let  $f: [0, 1] \times [0, 1] \rightarrow \mathbb{R}_+$  be a continuous function and let

$$m(x, B) = \int_B f(x, y) d\lambda(y),$$

where  $\lambda$  denotes the Lebesgue measure on  $[0, 1]$ . Then  $m$  is a transition measure.

## Продолжение Примера 2.4.16

(c)  $P = (p_{ij})_{i,j=1}^n$  is an  $n \times n$  matrix with nonnegative entries (i.e.  $p_{ij} \geq 0$  for all  $1 \leq i, j \leq n$ ) and

$$\sum_{j=1}^n p_{ij} = 1$$

(i.e. each row adds up to 1). Let

$$\Omega_1 = \Omega_2 = \{1, 2, \dots, n\}$$

and for any  $k \in \Omega_1$  and  $B \subseteq \Omega_2$  set

$$m(k, B) = \sum_{j \in B} p_{kj}.$$

Then  $m$  is a transition probability and the matrix  $P$  is called the associated transition probability matrix.

PROPOSITION 2.4.17 If  $(\Omega_1, \Sigma_1)$  and  $(\Omega_2, \Sigma_2)$  are measurable spaces,

$$f: \Omega_1 \times \Omega_2 \rightarrow \overline{\mathbb{R}}_+ = \mathbb{R}_+ \cup \{+\infty\}$$

is a  $\Sigma_1 \times \Sigma_2$ -measurable function and  $m(\cdot, \cdot)$  is a  $\sigma$ -finite transition measure on  $\Omega_1 \times \Sigma_2$ , then

$$\omega_1 \mapsto \int_{\Omega_2} f(\omega_1, \omega_2) m(\omega_1, d\omega_2)$$

is a  $\Sigma_1$ -measurable function from  $\Omega_1$  to  $\overline{\mathbb{R}}_+$ .

DEFINITION 2.4.18 Let  $(\Omega_1, \Sigma_1)$  and  $(\Omega_2, \Sigma_2)$  be measurable spaces. A  $\sigma$ -finite transition measure on  $(\Omega_1, \Sigma_2)$   $m$  is said to be "uniformly  $\sigma$ -finite", if there exist sequences

$$\{B_k\}_{k \geq 1} \subseteq \Sigma_1$$

and

$$\{D_n\}_{n \geq 1} \subseteq \Sigma_2$$

such that

$$\bigcup_{k \geq 1} B_k = \Omega_1, \quad \bigcup_{n \geq 1} D_n = \Omega_2$$

and

$$\sup_{\omega_1 \in B_k} m(\omega_1, D_n) < +\infty$$

for all  $k, n \geq 1$ .



PROPOSITION 2.4.19 If  $(\Omega_1, \Sigma_1)$  and  $(\Omega_2, \Sigma_2)$  are measurable spaces,  $m$  is a uniformly  $\sigma$ -finite transition measure on  $(\Omega_1, \Sigma_2)$ ,  $\mu$  is  $\sigma$ -finite measure on  $(\Omega_1, \Sigma_1)$  and  $C \in \Sigma_1 \times \Sigma_2$ , then

$$\nu(C) = \int_{\Omega_1} \left( \int_{\Omega_2} \chi_C(\omega_1, \omega_2) m(\omega_1, d\omega_2) \right) d\mu$$

is well defined and it is a  $\sigma$ -finite measure on  $\Sigma_1 \times \Sigma_2$ .

Moreover, if

$$f: \Omega_1 \times \Omega_2 \rightarrow \overline{\mathbb{R}}_+ = \mathbb{R}_+ \cup \{+\infty\}$$

is a  $\Sigma_1 \times \Sigma_2$ -measurable, then

$$\int_{\Omega_1 \times \Omega_2} f d\nu = \int_{\Omega_1} \left( \int_{\Omega_2} f(\omega_1, \omega_2) m(\omega_1, d\omega_2) \right) d\mu.$$

REMARK 2.4.20 If  $m(\omega_1, B) = m(B)$  (i.e. is independent of  $\omega_1$ , with  $m$   $\sigma$ -finite measure on  $\Sigma_2$ , then the  $\sigma$ -finite measure  $\nu$  established in Proposition 2.4.19 has the property that

$$\nu(A \times B) = \mu(A)m(B)$$

for all  $A \in \Sigma_1$  and  $B \in \Sigma_2$ ,  
i.e.  $\nu$  is the product of  $\mu$  and  $m$   
(see Definition 8.4.7).

Now we can state the generalized version of the Fubini-Tonelli theorem.

## THEOREM 2.4.21 (Generalized Fubini-Tonelli Theorem)

If  $(\Omega_1, \Sigma_1)$  and  $(\Omega_2, \Sigma_2)$  are measurable spaces,  $\mu$  is a  $\sigma$ -finite measure on  $\Sigma_1$ ,  $m$  is a uniformly  $\sigma$ -finite transition measure on  $\Omega_1 \times \Sigma_2$ ,  $\nu$  is the  $\sigma$ -finite measure on  $\Sigma_1 \times \Sigma_2$  obtained in Proposition 2.4.19 and  $f \in L^1(\Omega_1 \times \Omega_2, \nu)$ , then

$$\int_{\Omega_2} |f(\omega_1, \omega_2)| m(\omega_1, d\omega_2) < +\infty$$

$\mu$ -a.e. on  $\Omega_1$ ;

$$\int_{\Omega_1} \left( \int_{\Omega_2} |f(\omega_1, \omega_2)| m(\omega_1, d\omega_2) \right) d\mu < +\infty$$

and

$$\int_{\Omega_1 \times \Omega_2} f d\nu = \int_{\Omega_1} \left( \int_{\Omega_2} f(\omega_1, \omega_2) m(\omega_1, d\omega_2) \right) d\mu.$$

REMARK 2.4.22 If  $m(\omega_1, B) = m(B)$  (i.e.  $m$  is independent of  $\omega_1 \in \Omega_1$ ) and  $m$  is  $\sigma$ -finite measure on  $\Sigma_2$ , then Theorem 2.4.21 reduces to the classical Fubini-Tonelli theorem (Theorem 2.4.10) .

The foregoing definitions and properties readily extend to any finite number of sets and measurable spaces.

In the infinite case some of the definitions have to be modified in order to preserve these properties (compare with the corresponding topological situation, see Section 1.2).

DEFINITION 2.4.23 Let  $\{(\Omega_n, \Sigma_n)\}_{n \geq 1}$  be measurable spaces. Set

$$\Omega = \prod_{n \geq 1} \Omega_n.$$

If

$$B_k \subseteq \prod_{n=1}^k \Omega_n,$$

then the set

$$B_k \times \prod_{n \geq k+1} \Omega_n$$

is called a "cylinder" with base  $B_k$ .

The cylinder is said to be "measurable" if

$$B_k \in \prod_{n=1}^k \Sigma_n.$$

## Продолжение Определения 2.4.23

If

$$B_k = \prod_{n=1}^k A_n$$

with  $A_n \in \Sigma_n$  for  $n = 1, 2, \dots, k$ , then  $B_k$  is said to be a "product measurable cylinder" or "interval" in  $\Omega$  with sides

$$\{A_n\}_{n=1}^k.$$

The minimal  $\sigma$ -field over the measurable cylinders is called the "product" of the  $\sigma$ -fields  $\{\Sigma_n\}_{n \geq 1}$  and it is denoted by

$$\prod_{n \geq 1} \Sigma_n.$$

## REMARK 2.4.24

$$\prod_{n \geq 1} \Sigma_n.$$

is also the minimal  $\sigma$ -field over the product measurable cylinders.  
If all  $\Sigma_n$  coincide with a fixed  $\sigma$ -field  $\Sigma$ , then

$$\prod_{n \geq 1} \Sigma_n.$$

is denoted by

$$\Sigma^\infty.$$

Also if all  $\Omega_n$  coincide with a fixed set  $\Omega$ , then

$$\prod_{n \geq 1} \Omega_n.$$

is denoted by

$$\Omega^\infty.$$

In this setting the classical product measure theorem (Theorem 2.4.6) extends as follows (for a proof we refer to Dudley (1989), p. 201).

THEOREM 2.4.25 If  $\{(\Omega_n, \Sigma_n, \mu_n)\}_{n \geq 1}$  are probability spaces,

$$\Omega = \prod_{n \geq 1} \Omega_n \quad \text{and} \quad \Sigma = \prod_{n \geq 1} \Sigma_n,$$

then there exists a unique probability measure  $\mu$  on  $\Sigma$  such that

$$\mu(\{\omega = (\omega_k)_{k \geq 1} \in \Omega : \omega_1 \in A_1, \dots, \omega_n \in A_n\}) = \prod_{k=1}^n \mu_k(A_k)$$

for all  $n \geq 1$  and all  $A_k \in \Sigma_k$ ,  $k = 1, 2, \dots$

We call  $\mu$  the product of the  $\mu_n$  and write

$$\mu = \prod_{n \geq 1} \mu_n.$$



Thus far our approach has been to introduce integration using the concept of measure.

Now we will reverse this process and given an "integral" operation with some suitable properties we will show that it can be represented as the integral with respect to some measure.

This integral is known as the "Daniell Integral".

So let  $V$  be a family of  $\mathbb{R}$ -valued functions on some set  $\Omega$ .

We assume that  $V$  is a vector lattice, i.e. a vector space (i.e. if  $f, g \in V$  and  $c \in \mathbb{R}$ , then  $cf + g \in V$ ) and if  $f, g \in V$ , then

$$f \vee g = \max\{f, g\} \in V.$$

Note that this implies that we have also that for any  $f, g \in V$ ,

$$f \wedge g = \min\{f, g\} \in V.$$

Indeed observe that

$$f \wedge g = -((-f) \vee (-g)).$$

EXAMPLE 2.4.26 For any measure space  $(\Omega, \Sigma, \mu)$ ,  $L^p(\Omega)$ ,  $(1 \leq p \leq \infty)$  is a vector lattice.

Also if  $X$  is a topological space, the space  $C_b(X)$  of all bounded, continuous  $\mathbb{R}$ -valued functions on  $X$  is a vector lattice.

However,  $C^1(\mathbb{R}) = \{f: \mathbb{R} \rightarrow \mathbb{R} : f \text{ is continuously differentiable}\}$  is a vector space but not a vector lattice.

DEFINITION 2.4.27 Given a set  $\Omega$  and a vector lattice  $V$  of real functions on  $\Omega$ , a "Daniell functional" or "Daniell integral", is a function  $I: V \rightarrow \mathbb{R}$  such that

(a)  $I$  is a linear, i.e.  $I(cf + g) = cI(f) + I(g)$  for all  $f, g \in V$  and all  $c \in \mathbb{R}$ ;

(b)  $I$  is nonnegative, i.e. if  $f \in V$  and  $f(\omega) \geq 0$  for all  $\omega \in \Omega$ , then  $I(f) \geq 0$ ;

(c)  $I(f_n) \downarrow 0$  whenever  $f_n \in V$  and  $f_n(\omega) \downarrow 0$  for all  $\omega \in \Omega$  as  $n \rightarrow \infty$ .

REMARK 2.4.28 Condition (c) is clearly equivalent to each of the following conditions:

(c1) if  $\{f_n\}_{n \geq 1} \subseteq V$  is an increasing sequence and there exists a function  $f \in V$  such that

$$f \leq \lim_{n \rightarrow \infty} f_n,$$

then

$$I(f) \leq \lim_{n \rightarrow \infty} I(f_n),$$

and

(c2) if  $\{f_n\}_{n \geq 1} \subseteq V$  are nonnegative functions and  $g \in V$  such that

$$g \leq \sum_{n \geq 1} f_n,$$

then

$$I(g) \leq \sum I(f_n).$$

EXAMPLE 2.4.29 Let  $V = C([0, 1])$  and  $I: V \rightarrow \mathbb{R}$  is the classical Riemann integral.

Clearly conditions (a) and (b) of Definition 2.4.27 are satisfied and condition (c) follows from Dini's theorem (see Theorem 1.6.28) and the properties of the Riemann integral.

Here is the main theorem to this alternative approach to integration. For a proof, we refer to Dudley (1989), p. 110 or Royden (1968), p. 297–299.

**THEOREM 2.4.30 (Daniell-Stone Theorem)**

If  $V$  is a vector lattice of  $\mathbb{R}$ -valued functions on a set  $\Omega$  such that  $f \wedge 1 \in V$  for all  $f \in V$  and  $I$  is a Daniell integral on  $V$ , then there is a  $\sigma$ -field  $\Sigma$  of subsets of  $\Omega$  and a measure  $\mu$  on  $\Sigma$  such that

$$I(f) = \int_{\Omega} f \, d\mu,$$

for all  $f \in V$ .

PROPOSITION 2.4.31 If  $V$  is a vector lattice of  $\mathbb{R}$ -valued functions on a set  $\Omega$ ,  $1 \in V$  and  $\Sigma$  is the smallest  $\sigma$ -field of subsets of  $\Omega$  such that each  $f \in V$  is  $\sigma$ -measurable, then for every Daniell functional  $I$  there is a unique measure  $\mu$  on  $\Sigma$  such that

$$I(f) = \int_{\Omega} f \, d\mu,$$

for all  $f \in V$ .

REMARK 2.4.32 In general constant functions need not belong to the vector lattice  $V$ .

Any vector lattice  $V$  containing the constant functions is called a "Stone vector lattice".