# **Contents**

1	From Reading Group on Durrett			2
	1.1	Measure and Integration		2
		1.1.1	Measure and Probability Spaces	2
		1.1.2	Construction of Measures	4
		1.1.3	Application: Borel and Lebesgue-Stietjes Measure on ${\mathbb R}$	6
		1.1.4	Integration	6
		1.1.5	Countable Product Space and Lebesgue Measure on $\mathbb{R}^d$	8
		1.1.6	Convergence Concepts	8
		1.1.7	Differentiation	8
	1.2	Indepe	endence	8
		1.2.1	Independence	8
		1.2.2	Law of Large Numbers	8
		1.2.3	Central Limit Theorem	8
Color code: highlight, key concepts and definitions, to be added.				

# **Chapter 1**

# From Reading Group on Durrett

Reference: PTE, Folland and Tao.

### 1.1 Measure and Integration

### 1.1.1 Measure and Probability Spaces

First of all, why should we discuss measure theory. It's for a better integration theory than the usual Riemann integral. Because we need strong conditions for arguments like:

$$\lim_{n} \int f_n = \int \lim_{n} f_n$$

to hold for Riemann integral. TBA: Relationship with Riemann Integral We'll talk about

- 1. Measure space, construction of measure, Lebesgue measures;
- 2. Integration theory
- 3. Convergence concepts
- 4. Differentiation

Let's start with an abstract measure space. Let  $\Omega$  be a space. A  $\sigma$ -algebra  $\mathcal{F}$  is a collection of subsets called *measurable sets* of  $\Omega$  such that it's closed under complements and countable unions. Also the  $\emptyset \in \mathcal{F}$ . We can assign *measure*  $\mu$  to members in  $\mathcal{F}$ . The

reason we need to restrict attention from the power set to  $\mathcal{F}$  is that there are examples that violate some properties we want a measure to have.  $\mu$  satisfies:

$$\mu(\varnothing) = 0$$
 and  $\mu\left(\bigsqcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} \mu(A_i)$  for  $(A_i) \subset \mathcal{F}$ 

The triplet  $(\Omega, \mathcal{F}, \mu)$  is called a *measure space*.

A *measurable function* is  $X:(\Omega,\mathcal{F},\mu)\to (S,\mathcal{H})$  such that for all  $H\in\mathcal{H},X^{-1}(H)\in\mathcal{F}$ . Meaning that we can assign a measure to the subsets of  $\mathcal{H}$  based on the measure  $\mu$  and X. Given two measurable spaces, we can check whether a function X is measurable.

**1.1.1** Lemma. Suppose  $\mathcal{H} = \sigma(\mathcal{G})$ , if for all  $B \in \mathcal{G}$ ,  $X^{-1}(B) \in \mathcal{F}$ , then X is measurable.

*Proof.* Let  $\mathcal{H}' = \{B \subset S \mid X^{-1}(B) \in \mathcal{F}\}$ , then we can show  $\mathcal{H}'$  is a  $\sigma$ -algebra because  $X^{-1}$  preserves complements and union. Then we have  $\mathcal{G} \subset \sigma(\mathcal{G}) = \mathcal{H} \subset \mathcal{H}'$ , meaning that all  $H \in \mathcal{H}$  satisfies the condition.

A measure space is a *probability space* if  $\mu = 1$ , we write it as  $(\Omega, \mathcal{F}, P)$ . A random variable/vector is a measurable function on probability space  $X : \Omega \to \mathbb{R}/\mathbb{R}^d$  with the Borel  $\sigma$ -algebra.

#### **Distribution and Density**

*Distribution* is the probability induced by a random variable X, such that  $\mu(A) = P(X \in A)$ . *Density* is defined via some methods(To be added). Its use is:

$$\int g(X) \, \mathrm{d}P = \int g(x) f(x) \, \mathrm{d}x$$

#### Check for Measurability of Random Variables

For random variables  $X_n$ ,  $Y_n$ ,  $X_n + Y_n$  and  $\sup X_n$  are measurable.

#### 1.1.2 Construction of Measures

#### Motivation

Now we want to show that we can actually find measure on  $\mathbb{R}$  and  $\mathbb{R}^d$  that are the most useful spaces. We start from abstract construction method inspired by the following observation of measure on  $\mathbb{R}$ .

In order to have a natural measure  $\lambda$  on  $\mathbb{R}$ , we want the following properties:

- 1.  $\lambda_0(a, b] = b a$ ;
- 2. Measure of union of intervals should be the sum;(we can't have uncountable sum, hence restricted to countable union)
- 3. Apply to all intervals(we can't apply to the power set, hence restrict to the measurable sets).

Now consider the set  $S_{\mathbb{R}} = \{(a, b] : a \leq b \in \mathbb{R}\}$ , it's too small, we want to extend the natural measure funtion  $\lambda_0$  to a larger set(we can extend to  $\mathcal{B}_{\mathbb{R}}$ , even to  $\mathcal{L}_{\mathbb{R}}$ ).

Semi-algebra S is collection of subsets of a space  $\Omega$  that satisfies the following conditions: closed under intersection and the complements are union of finite disjoint sets in S. Let  $\mu_0$  be a what I call semi-pre-measure to (algebra, premeasure) to (outer-measurable sets, outer measure) to (sigma-algebra, measure) where the semi-pre-measure  $\mu_0$  satisfies if both  $S_N = \bigsqcup^N S_i, S_\infty = \bigsqcup^\infty S_i \in S$ 

$$\mu_0(\emptyset) = 0; \quad \mu_0(S_N) = \sum_{j=0}^{N} \mu_0 S_j; \quad \mu_0(S_\infty) \le \sum_{j=0}^{\infty} \mu_0 S_j$$

We follow the steps of extension

$$(S, \mu_0) \to (A, \mu_1) \to (\mathcal{M}, \mu^*) \to (\sigma(S), \mu)$$

that is semi-algebra and *pre-measure*  $\mu_1$  satisfies  $\mu_1 \bigsqcup_{i=1}^{\infty} A_i = \sum_{i=1}^{\infty} \mu_1 A_i$  as long as the union is also in  $\mathcal{A}$ . And we will show that  $\mathcal{B}_{\mathbb{R}} \subset \sigma(\mathcal{S})$ .

#### **Construction of Abstract Measure**

**1.1.2**  $(S, \mu_0) \to (A, \mu_1)$ . We have that  $A = \{ \bigsqcup_j^N S_j : S_j \in S \}$  is an algebra. And  $\mu_1(A) = \sum_j^N \mu_0 S_j$  is a premeasure.

*Proof.*  $\mu_1(\emptyset) = 0$ . Suppose  $A = \bigsqcup^{\infty} A_i \in \mathcal{A}$ , then there exists  $T_l \in \mathcal{S} : l = 1, \dots N_T$ , such that  $A = \bigsqcup_l T_l$ , also for each  $A_j = \bigsqcup_k^{N_j} S_{j,k}$ .

Let an *outer measure* be a function over the power set such that:

$$\mu^*(\varnothing) = 0$$
 and  $\mu^*\left(\bigsqcup^{\infty} A_j\right) \le \sum_j \mu^* A_j$  and  $\mu^*(A) \le \mu^*(B)$  for  $A \subset B$ 

Also let M be the *outer-measurable sets*, that is the collections of A such that

$$\mu^*(E) = \mu^*(E \cap A) + \mu^*(E \cap A^c)$$
 for all  $E \subset X$ 

1.1.3  $(A, \mu_1) \to (M, \mu^*)$  Caratheodory's Theorem applied to premeasures. We can find the outer measure induced by  $\mu_1$ :

$$\mu_1^*(B) = \inf \left\{ \sum_{j=1}^{\infty} \mu_1 A_j : B \subset \bigcup_{j=1}^{\infty} A_j \right\}$$

we show  $\mu_1^*$  is indeed an outer measure, Caratheodory states that for outer measures, M of  $\mu_1^*$  is a  $\sigma$ -algebra and  $\mu^*|_{\mathcal{M}}$  is complete.

Proof.

1. Monotone property is easy, for countable subadditivity, we use  $\varepsilon$ -room method, find a cover  $C_{jk}$  of  $A_j$  such that :

$$\mu_1^*(A_j) > \sum_k \mu_1^* C_{jk} - \varepsilon$$

2. To show that  $\mathcal{M}$  is a *σ*-algebra we only need to show that for  $E_j \in \mathcal{M}$  and any  $B \subset \Omega$ , we have:

$$\mu^*(B) \ge \mu^*(B \cap E) + \mu^*(B \cap E)$$

which can be seen by considering finite sum, taking out one  $E_j$  at a time and then take limit.

- 3. Show that the restriction is complete.
- **1.1.4**  $(\mathcal{M}, \mu^*) \to (\sigma(S), \mu)$ . We can restrict  $\mu^*$  to define  $\mu = \mu^*|_{\sigma(S)}$ , and the resulting measure is complete.

*Proof.* We show that  $\sigma(S) \subset M$  and the restriction is complete.

### 1.1.3 Application: Borel and Lebesgue-Stietjes Measure on R

We verify that for any nondecreasing right-continuous function  $F : \mathbb{R} \to \mathbb{R}$ , the semi-algebra  $\mathcal{S}_{\mathbb{R}}$  together with the semi-pre-measure defined by

$$\lambda_0(a,b] = F(b) - F(a)$$

satisfies the conditions set in the previous section.  $\mathcal{A}_{\mathbb{R}}$  be the algebra generated by  $\mathcal{S}_{\mathbb{R}}$  in turn will generate  $\mathcal{B}_{\mathbb{R}}$ . We can extend it to a something larger, a complete measure whose domain contains  $\mathcal{B}_{\mathbb{R}}$ . It will be called the *Lebesgue-Stietjes measure* and *Lebesgue measurable sets*  $\mathcal{M}_{\lambda}$ . The restriction on  $\mathcal{B}$  is called *Borel measure*. It's just the completion.

- **1.1.5** Regularity of  $\mathcal{M}_{\lambda}$ . Lebesgue measurable sets are of simple form if you allow for a small error.
  - 1.  $\lambda(A) = \inf \{ \lambda U : A \subset U, U \text{ is open} \} = \sup \{ \lambda K : K \subset A, K \text{ is compact} \}$
  - 2.  $A = G_{\delta} \setminus N_1 = F_{\sigma} \cup N_2$  where N are null sets.
  - 3. Littlewood's First Principle: Suppose  $\lambda(A) < \infty$  then for all  $\varepsilon > 0$ , there exists B that is a finite union of open intervals such that  $A \triangle B < \varepsilon$ .

*Proof.* 

#### 1.1.4 Integration

Integration and measure are the two sides of same coin: integration is the generalization of measure from the space of *indicator functions*  $1_A$  of measurable sets to a larger set of functions.

For *simple functions*  $f_s = \sum_{j=1}^{N} \mathbf{1}_{A_j}$ , we can define  $\mu f_s = \sum_{j=1}^{N} \mu(A_j)$ Then for *nonnegative measurable* functions f, we can define

$$\mu f = \sup \{ \mu f_s : f_s \le f \text{ pointwise} \}$$

If  $f = f^+ - f^-$  and the integrals of two nonnegative parts are not  $\infty$ , then we say f is *integrable* or in  $L^1$  and  $\mu f = \mu f^+ - \mu f^-$ .

- **1.1.6 Littlewood's Second Principle.** Measurable and integrable functions can be approximated by simple well-behaved functions.
- **1.1.7** Properties of Integral. 1. Monotone  $\mu f \leq \mu g$  if  $f \leq g$  pointwise.
  - 2. Linearity for a > 0,  $\mu(af) = a\mu f$ ,  $\mu(f+g) = \mu f + \mu g$ .
  - 3. Monoton Convergence: if  $0 \le f_n \uparrow f$  a.e., then  $\mu f_n \uparrow \mu f$ .

*Proof.* Folland proves by choosing simple function and a scaling factor  $\alpha$ . Tao's proof is essentially the same. PTE proves via Fatou's lemma.

Suppose  $f_n \uparrow f$  a.e., if  $\mu f$  exists, we know  $\mu f_n$  must converge because it's a nondecreasing bounded sequence of numbers. Also  $\mu f_n \leq \mu f$  by monotonicity.

To show that  $\mu f \leq \lim_n \mu f_n$ , where  $\mu f = \sup \{\mu g : g \text{ is simple, } g \leq f\}$ , we only need to show that for all simple  $g \leq f$ , we have  $\mu g \leq \mu f_n$ . We can use  $\varepsilon$ -room method, let  $E_n = \{\omega : f_n \geq (1 - \varepsilon)g\}$ ,  $E_n \uparrow \Omega$ , and we have  $\int f_n \geq \int_{E_n} (1 - \varepsilon)g$ , taking  $\lim_n f_n \leq (1 - \varepsilon) \int g$  for all  $\varepsilon > 0$ .

*Remark*. These three properties fully characterise the integral. That is given a functional that satisfies the three properties we can find a measure for which the function is the integral.

- 1.1.8 Inequalities and Controls. Useful inequalities and controls
  - 1. Jensen's inequality: suppose we have a convex function  $\Phi(x)$ , then  $\mu(\Phi(f)) \geq \Phi \mu f$
  - 2. Holder's inequality. For  $p, q : \frac{1}{p} + \frac{1}{q} = 1$ , we have  $||fg||_1 \le ||f||_p ||g||_q$
  - 3. Minkowski's inequality:  $||f + g||_p \le ||f||_p + ||g||_p$ .
  - 4. Chebyshev's Inequality:
  - 5. For nonnegative measurable functions,  $\mu f < \infty \implies f < \infty$  a.e.,  $\mu f = 0$  implies f = 0 a.e.
  - 6. Moment and tail behaviour, if  $\mu$  is finite,

$$\mu f^r < \infty \implies \sum x^{r-1} \mu(f > x) < \infty$$

## 1.1.5 Countable Product Space and Lebesgue Measure on $\mathbb{R}^d$

Product measurable and product measure.

- 1.1.9 Fubini-Tonelli.
  - 1.1.6 Convergence Concepts
- 1.1.10 Monotone, Fatou and Dominated Convergence.
- 1.1.11 Uniform Integrability.
  - 1.1.7 Differentiation
  - 1.2 Independence
  - 1.2.1 Independence
  - 1.2.2 Law of Large Numbers
  - 1.2.3 Central Limit Theorem