## STOCHASTIC CALCULUS

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## Chapter 1

### **Probabilities**

#### 1.1 Event Sets

<u>DEFINITION</u>. A collection (or call it a set)  $\mathcal{F}$  of subsets of  $\Omega$  is called a  $\sigma$ -algebra if it satisfies:

- 1. contains the empty set:  $\emptyset \in \mathcal{F}$ ;
- 2. is closed under **countable** unions:  $A_1, A_2, ..., \in \mathcal{F} \implies \bigcup_i A_i \in \mathcal{F}$ ;
- 3. is closed under complements:  $A \in \mathcal{F} \implies A^C \in \mathcal{F}$ ;

It is trivial to know:

$$A_1, A_2, \dots \in \mathcal{F} \implies \bigcup_i A_i^C \in \mathcal{F}$$
  
 $\implies (\bigcup_i A_i^C)^C \in \mathcal{F}$   
 $\implies \cap_i A_i \in \mathcal{F}$ 

<u>Trival  $\sigma$ -algebra</u>.  $\mathcal{F}_0 = \{\emptyset, \Omega\}$  is a  $\sigma$ -algebra.

Collection of all subsets.  $\mathcal{F} = 2^{\Omega}$  is a  $\sigma$ -algebra.

<u>DEFINITION</u>. Let  $\mathcal{G}$  be a collection of subsets of  $\Omega$ . The  $\sigma$ -algebra generated by  $\mathcal{G}$ :  $\sigma(\mathcal{G})$  is the **smallest**  $\sigma$ -algebra that contains  $\mathcal{G}$ .

<u>DEFINITION</u>. A pair is called a *measure space*  $(\Omega, \mathcal{F})$  if  $\Omega$  is the sample space and  $\mathcal{F}$  is a  $\sigma$ -algebra of subsets.

### 1.2 Probability

<u>DEFINITION</u>. A function  $\mathbb{P}$  defined on  $(\Omega, \mathcal{F})$ :  $\mathbb{P} : \mathcal{F} \to [0, 1]$  is called a *probability measure* if:

- 1.  $\mathbb{P}(\Omega) = 1$
- 2. Only for *countable* unions: if  $A_i \cap A_j = \emptyset$  for  $i \neq j \implies \mathbb{P}(\cup_i A_i) = \sum_i \mathbb{P}(A_i)$

The countable sample space is easy to handle with, take  $\mathcal{F} = 2^{\Omega}$  and some  $\mathbb{P}$  to assign each event to [0, 1]. But the uncountable space could be more delicate to solve.

### 1.3 Infinite Spaces

### 1.3.1 Uniform Lebesgue Measure on (0, 1)

Define a Lebesgue measure  $\mu$ , then we can determine the probability that  $\omega$  falls within an **open interval**:

$$\mathbb{P}(\{\omega : \omega \in (a,b)\}) = \mu((a,b)) := b - a, 0 < a \le b < 1$$

In this case we notice that  $2^{\Omega} = 2^{(0,1)}$  is not a  $\sigma$ -algebra of (0, 1). Then we introduce the *Borel*  $\sigma$ -algebra on (0, 1) to make an appropriate sample space for our experiment:

$$\mathcal{B}((0,1)) := \sigma(\mathcal{O}) \text{ where } \mathcal{O} = \{ A \subseteq (0,1) : A = (a,b), 0 < a \le b < 1 \}$$

This note will not discuss much on the *Borel set*, it is about creating some subsets of open sets in  $\Omega$ .

#### 1.3.2 Infinite Sequence of Coin Tosses

Let  $\omega = \omega_1 \omega_2 ... \omega_n$  where  $\omega_i \in \{H, T\}$ . If I have known  $\omega_1, \omega_2$ , I can tell you if  $\omega$  belongs to each of the sets in  $\mathcal{F}_2$ : all possible cases for two tossing. When n becomes very large, then we have:

$$\mathcal{F} = \sigma(\mathcal{F}_{\infty}), \ \mathcal{F}_{\infty} = \cup_{\infty} \mathcal{F}_{n}$$

But things (or call them sets) like "sequences for which x percent of coin tosses are heads" are not in  $\mathcal{F}_{\infty}$ , they are actually in  $\mathcal{F}$ .

<u>DEFINITION</u>. Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space, then if a set  $A \in \mathcal{F}$  s.t.  $\mathbb{P}(A) = 1$ , then the event A occurs  $\mathbb{P}$  almost surely (i.e.  $\mathbb{P}$ -a.s.).

# Chapter 2

# Information and Conditioning

### 2.1 Information and $\sigma$ -Algebras

<u>DEFINITION</u> 2.1.1. Let  $\Omega$  be a nonempty set. Let T be a fixed positive number, and assume that for each  $t \in [0, T]$  there is a