# Math 418 Notes

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1 Introduction

(ii)  $\mu(\dot{\bigcup}_{i\in I}A_i) = \sum_{i\in I}\mu(A_i)$ 

where  $I = \mathbb{N}$  or  $I = \{1, \dots, n\}$ .

a  $probability\ space.$ 

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1 Introduction	
Text: R. Durrett. Probability Theory adn Examples. Version 5a.	
<b>Definition 1.1.</b> A sample space $\Omega$ is a non-empty set. The set of subsets $\mathcal{F} \in 2^{\Omega}$ is a field if and only if	
$\text{(i)} \ \ \emptyset \in \mathcal{F}.$	
(ii) Closed under complement. $\forall A \in \mathcal{F}, A^C \in \mathcal{F}.$	
(iii) Closed under finite union. $\forall A, B \in \mathcal{F}, A \cup B \in \mathcal{F}.$	
<b>Definition 1.2.</b> The set of subsets $\mathcal{F} \in 2^{\Omega}$ is a $\sigma$ -field iff	
$\text{(i)} \ \ \emptyset \in \mathcal{F}.$	
(ii) Closed under complement. $\forall A \in \mathcal{F}, A^C \in \mathcal{F}.$	
(iii) Closed under countable union.	
<b>Definition 1.3.</b> Call $(\Omega, \mathcal{F})$ a measureable space if $\mathcal{F}$ is a $\sigma$ -field.	
<b>Definition 1.4.</b> We write $A = \dot{\bigcup}_{i \in I} A_i$ if the sets $A_i$ are disjoint.	
<b>Definition 1.5.</b> Let $(\Omega, \mathcal{F})$ be a measurable space. The measure $\mu$ on $(\Omega, \mathcal{F})$ is a countab additive, non-negative set function $\mu : \mathcal{F} \to [0, \infty]$ such that	oly
(i) $\mu(\emptyset) = 0$	

 $(\Omega, \mathcal{F}, \mu)$  is a measure space. If  $\mu$  is a probability, that is  $\mu(\Omega) = 1$ , then the said measure space is

**Definition 1.6.** A finite additive measure on  $(\Omega, \mathcal{F})$ , where  $\mathcal{F}$  is a field, is a function  $\mu : \mathcal{F} \to [0, \infty]$  such that

- (i)  $\mu(\emptyset) = 0$
- (ii)  $\mu(\dot{\bigcup}_{i\in I}A_i) = \sum_{i\in I}\mu(A_i)$

where (ii) only holds for I finite.

#### **Example 1.1.** An example of a finite additive measure is

$$\mu(A) = \begin{cases} 0 \text{ if A is finite} \\ \infty \text{ if A is infinite} \end{cases}$$

**Lemma 1.1.** Let  $\mathcal{F}$  be a field and  $\mu$  be a finite additive measure, then

- (a) Monotonicity.  $A, B \in \mathcal{F}$  with  $A \subset B$  implies  $\mu(A) \leq \mu(B)$ .
- (b) Subadditivity.  $A, B \in \mathcal{F}$  means  $\mu(A \cup B) \leq \mu(A) + \mu(B)$ .

Proof of (a). Elementary.

*Proof of (b).* Use (a) and disjoint union.

**Definition 1.7.** We write  $A_n \uparrow A$  if  $A_n \subset A_{n+1}$  and  $\bigcup A_n = A$ . Defined similarly for  $\downarrow$ .

**Theorem 1.2.** Let  $\mu$  be a measure on  $(\Omega, \mathcal{F})$ , and  $\{A_n : n \in \mathbb{N}\}$  be a sequence in  $\mathcal{F}$ .

- (a) Continuity from below.  $A_n \uparrow A \implies \mu(A_n) \uparrow \mu(A)$
- (b) Continuity from above.  $\mu(A_1) < \infty$  and  $A_n \downarrow A \implies \mu(A_n) \downarrow \mu(A)$
- (c) Countable subadditivity.  $\mu(\bigcup_{n=0}^{\infty} A_n) \leq \sum_{n=0}^{\infty} \mu(A_n)$

Proof of (a). Observe that  $A_n = \bigcup_{k=1}^n (A_k \setminus A_{k-1})$  where we define  $A_0 = \emptyset$ . Then

$$\mu(A) = \sum_{k=1}^{\infty} \mu(A_k \setminus A_{k-1}) = \lim_{n \to \infty} \mu(A_n).$$

Also,  $\mu(A_n) \leq \mu(A_{n+1})$ , so  $\mu(A_n) \uparrow \mu(A)$ .

*Proof of (b).* We can take complements with respect to  $A_1$  and use (a) on  $(A_1 \setminus A_n) \uparrow (A_1 \setminus A)$ . Then  $\mu(A_1 \setminus A_n) \uparrow \mu(A_1 \setminus A)$ .

*Proof of (c).* We use finite subadditivity as follows

$$\mu(\cup_k^n A_k) \le \sum_k^n \mu(A_k) \le \lim_{n \to \infty} \sum_k^n \mu(A_k).$$

Partial converse to Theorem 1.2 (Continuity from below): If  $\mu$  is finitely additive on  $(\Omega, \mathcal{F})$  and continuous from below, then  $\mu$  is a measure (countably additive).

**Example 1.2.** Consider the counting measure. Let  $A_n = \{n, n+1, \ldots\}$ , then  $A_n \downarrow \emptyset$ , but  $\mu(A_n) = \infty$  for all n and  $\mu(\emptyset) = 0$ .

**Definition 1.8.** A  $\mathcal{A} \subset 2^{\Omega}$  is a  $\pi$ -system if it is closed under intersection. A  $\mathcal{A} \subset 2^{\Omega}$  is a  $\lambda$ -system if

- (i)  $\Omega \in \mathcal{A}$
- (ii) If  $A, B \in \mathcal{A}$  and  $B \subset A$ , then  $A \setminus B \in \mathcal{A}$
- (iii) If  $A_n \in \mathcal{A}$  and  $A_n \uparrow A$ , then  $A \in \mathcal{A}$

**Theorem 1.3** ( $\pi$ - $\lambda$  Theorem). Let  $\mathcal{C}$  be a  $\pi$ -system in  $2^{\Omega}$ , and  $\mathcal{L}$  be a  $\lambda$ -system in  $2^{\Omega}$ . If  $\mathcal{C} \subset \mathcal{L}$ , then  $\sigma(\mathcal{C}) \subset \mathcal{L}$ .

*Proof.* Draw a venn diagram, or show that the smallest  $\lambda$ -system containing  $\mathcal{C}$  is a  $\sigma$ -field.

**Corollary 1.3.1.** Let  $P_1$  and  $P_2$  be probabilities on  $(\Omega, \mathcal{F})$ , and  $\mathcal{C} \subset \mathcal{F}$  be a  $\pi$ -system. If  $P_1 = P_2$  on  $\mathcal{C}$ , then  $P_1 = P_2$  on  $\sigma(\mathcal{C})$ .

*Proof.* Let  $\mathcal{L} = \{A \in \mathcal{F} : P_1(A) = P_2(A)\}$ , then  $\mathcal{C} \subset \mathcal{L}$ . We want to show that  $\mathcal{L}$  is a  $\lambda$ -system.

- (i) Clearly  $\Omega \in \mathcal{L}$ .
- (ii) If  $A, B \in \mathcal{L}$  and  $A \subset B$ , then  $B \setminus A \in \mathcal{L}$  by finite additivity.
- (iii) If  $A_n \in \mathcal{L}$  and  $A_n \uparrow A$ , then we can use continuity from below to get  $P_1(A) = P_2(A)$ .

**Example 1.3.** Let  $p_1, \ldots, p_6 \in [0, 1]$  and  $\sum_{i=1}^6 p_i = 1$ . We can define  $\Omega = \{1, \ldots, 6\}$ ,  $\mathcal{F} = 2^{\Omega}$ , and  $P(\{i\}) = p_i$ . Then by finite additivity, we get  $P(A) = P(\dot{\cup}_{i \in A} \{i\}) = \sum_{i \in A} p_i$  for any  $A \in \mathcal{F}$ .

## 2 The Law of Large Numbers

#### 2.1 The Law of Large Numbers for Coin Tossing

**Experiment**: Toss a fair coin an infinite number of times.

**Outcome**: An infinite sequence of 0's and 1's.  $w = (w_1, w_2, ...)$ .

Let  $\Omega = \{0,1\}^{\mathbb{N}}$  be the sample space. Let  $x_n : \Omega \to \{0,1\}$  with  $x_n(w) = w_n$ , and  $s_n : \Omega \to \{0,1,\ldots,n\}$  with  $s_n(w) = \sum_{i=1}^n x_i(w)$ .

**Definition 2.1.** The event  $A \subset \Omega$  is a finite dimensional event if  $\exists n, B \subset \{0,1\}^n$  such that  $A = \{w : (w_1, w_2, \dots, w_n) \in B\}$ . (ie. we only need to look at the first n things.) Let  $\mathcal{F}_0$  be the set of finite dimensional events.

We want to show that  $\lim_{n\to\infty} \frac{s_n(n)}{n} = \frac{1}{2}$ .

**Lemma 2.1.**  $\mathcal{F}_0$  is a field.

*Proof.* Observe that  $\emptyset = \{w \in \Omega : w_1 \in \emptyset\} \in \mathcal{F}_0$ .

Let  $A_1, A_{@} \in \mathcal{F}$ , then there exist  $n_1, n_2, B_1, B_2$  such that  $A_i = \{w : (w_1, \dots, w_{n_i}) \in B_i\}$  for  $i \in \{1, 2\}$ . Now we may assume without loss of generality that  $n_1 = n_2$ , because we can extend the smaller  $B_i$  so that it has the same dimension as the larger one. This is also closed under complement.

**Definition 2.2.** Define  $P: \mathcal{F}_0 \to [0,1]$  such that  $P(\{w: (w_1,\ldots,w_n) \in B\}) = \frac{\#B}{2^n}$ , where  $B \subset \{0,1\}^n$ . We can check that P is well defined.

Lemma 2.2. P is a probability.

*Proof.* We can check that  $P(\emptyset) = 0$  and  $P(\Omega) = 1$ .

Let disjoint  $A_1, A_2 \in \mathcal{F}$ . Then  $\emptyset = A_1 \cap A_2 = \{w : (w_1, \dots, w_n) \in B_1 \cap B_2\}$ , implying that  $B_1 \cap B_2 = \emptyset$ . Then, we can write the disjoint union in the same way.

We can compute  $P({s_n(w) = k}) = {n \choose k} 2^{-n}$ , and say " $s_n$  has binomial distribution with parameters  $(n, \frac{1}{2})$ ".

Let  $C = \left\{ w : \lim_{n \to \infty} \frac{s_n(w)}{n} = \frac{1}{2} \right\}$ . Observe that  $C \notin \mathcal{F}_0$ . However, we will extend P from  $\mathcal{F}_0$  to  $\sigma(\mathcal{F}_0)$ .

Lemma 2.3.  $C \in \sigma(\mathcal{F}_0)$ .

Proof. Observe that

$$w \in C \iff \lim_{n \to \infty} \frac{S_n(w)}{n} = \frac{1}{2}$$

$$\iff \forall M \in \mathbb{N}, \ \exists N \text{ such that } \forall n \ge N, \ \left| \frac{S_n(w)}{n} - \frac{1}{2} \right| < \frac{1}{M}$$

$$\iff w \in \bigcap_{M=1}^{\infty} \bigcup_{N=1}^{\infty} \bigcap_{n=N}^{\infty} \left\{ w : \left| \frac{S_n(w)}{n} - \frac{1}{2} \right| < \frac{1}{M} \right\}$$

Let  $C_{n,M}$  be the latter set, which is finite dimensional, hence  $C_{n,M} \in \mathcal{F}_0$ . It follows that  $C \in \sigma(\mathcal{F}_0)$  as  $\sigma$ -fields are closed under finite unions and intersections.

Suppose that we can extend P to  $\sigma(\mathcal{F}_0)$  (by Carathéodory). Now we need to prove that P(C) = 1, that is the event happens almost surely.

**Lemma 2.4.**  $P(C_{n,M}^c) \leq \frac{M^2}{4n}$ .

*Proof.* We observe that  $S_n$  is a binomial random variable, and  $\frac{n}{2}$  is the mean. Then we can use Chebychev's inequality.

$$P\left(\left|\frac{S_n}{n} - \frac{1}{2}\right| \ge \frac{1}{M}\right) = P\left(\left|S_n - \frac{n}{2}\right| \ge \frac{n}{M}\right) \le \frac{\operatorname{var}(S_n)}{(n/M)^2} = \frac{M^2}{4n}.$$

There is also a direct calculation proof given on Perkins's webpage.

**Theorem 2.5** (The Strong Law of Large Numbers for Coin Tossing). P(C) = 1.

*Proof.* Let  $\hat{C} = \left\{ w : \lim_{m \to \infty} \frac{S_{m^2}}{m^2} = \frac{1}{2} \right\}$ . This set is equal to C as shown in homework 1. Similarly, we can write

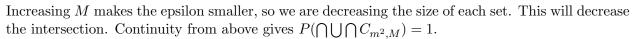
$$\hat{C} = \bigcap_{M=1}^{\infty} \bigcup_{N=1}^{\infty} \bigcap_{m=N}^{\infty} \left\{ \left| \frac{S_{m^2}}{m^2} - \frac{1}{2} \right| < \frac{1}{M} \right\}.$$

We want to show that  $P\left(\bigcap_{m=N}^{\infty} C_{m^2,M}\right) = 1$ , after which continuity from below and continuity from above will yield the result.

$$P\left(\bigcap_{m=N}^{\infty} C_{m^2,M}\right) = 1 - P\left(\bigcup_{m=N}^{\infty} C_{m^2,M}^c\right) \ge 1 - \sum_{m=N}^{\infty} P(C_{m^2,M}^c) \ge 1 - \sum_{m=N}^{\infty} \frac{M^2}{4m^2} \to 1,$$

where the convergence holds because  $\frac{1}{m^2}$  is summable.

Increasing N means we take the intersection of less sets, which increases. Continuity from below gives  $P(\bigcup \bigcap C_{m^2,M}) = 1$ .



Hence  $P(C) = P(\hat{C}) = 1$ .

Now we prove that we can indeed extend P to  $\sigma(\mathcal{F}_0)$ .

**Theorem 2.6.** Let C be a  $\pi$ -system,  $\mu_1, \mu_2$  be measures on  $\sigma(C)$ . If  $\mu_1 = \mu_2$  on C and both are  $\sigma$ -finite on C, then  $\mu_1 = \mu_2$  on  $\sigma(C)$ .

Proof. Let  $\{A_n\} \in \mathcal{C}$  and  $A_n \uparrow \Omega$  where  $\mu_1(A_n) = \mu_2(A_n) < \infty$ . For all  $\mu_i(A_n) > 0$ , define  $P_{i,n}(\mathcal{E}) = \frac{\mu_i(E \cap A_n)}{\mu_i(A_n)}$ . Then  $P_{i,n}$  is a probability. We can check that  $\mu_i(A_n) > 0$  for n sufficiently large. We can also show that  $P_{1,n} = P_{2,n}$  using continuity from below. Now we can use  $\pi$ - $\lambda$  theorem to finish the proof.

**Lemma 2.7.** Let  $P: \mathcal{F}_0 \to [0,1]$  be a finitely additive probability, then P is a countably additive probability, where  $\mathcal{F}_0$  is defined as above.

*Proof.* Give  $\Omega = \{0,1\}^{\mathbb{N}}$  the product topology. That is for  $\{w_n\} \subset \Omega$ , then  $w^n \to w$  as  $n \to \infty$  if and only if  $\forall i \in \mathbb{N}$ , we have  $\lim_{n \to \infty} w_{n,i} = w_i$ . Equivalently, we can define the metric  $d(w,w') = \sum_{i=1}^{\infty} \frac{|w_i - w_i'|}{2^i}$ . Observe that  $(\Omega, d)$  is compact by Cantor diagonalization.

 $\sum_{i=1}^{\infty} \frac{|w_i - w_i'|}{2^i}.$  Observe that  $(\Omega, d)$  is compact by Cantor diagonalization. Let  $\{A_n\}$  be disjoint sets in  $\mathcal{F}_0$  such that  $\bigcup_{n=1}^{\infty} A_n \in \mathcal{F}_0$ . Let  $K_n = \bigcup_{j=1}^{\infty} A_j \setminus \bigcup_{j=1}^n A_j$ . Observe that  $K_n \downarrow \emptyset$ . If each  $K_n$  is compact, then  $\exists N$  such that  $forall n \geq N$ , we have  $K_n = \emptyset$ . Hence finite additivity implies countable additivity.

Now it suffices to show that each  $K_n = \bigcup_{j=1}^{\infty} A_j \setminus \bigcup_{j=1}^n A_j$  is compact.