# ORFE 524: Statistical Theory and Methods Homework 1

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**Exercise 1:** Recall the definition of  $\sigma$ -Algebra. Let  $(\Omega, \Sigma)$  be a measurable space, that is,  $\Sigma$  satisfies the following three properties:

- 1.  $\Sigma \neq \emptyset$ ,  $\Sigma \subseteq 2^{\Omega}$ .
- 2.  $A \in \Sigma$  implies that  $A^c \in \Sigma$ . Here we use  $A^c$  to denote the complement of A.
- 3.  $\forall A_1, A_2, \ldots \in \Sigma$ , we have  $\cap_{i>1} A_i \in \Sigma$ .

Based on these properties, solve the following problems:

- 1) Show that  $\Sigma$  is closed under union.
- 2) Show that  $\Sigma$  must contain  $\emptyset$  and  $\Omega$ .
- 3) Suppose  $A \subseteq \Omega$ , what is the smallest  $\sigma$ -algebra containing A?
- 4) Show that the set of all rational numbers, denoted by  $\mathcal{Q}$ , is Borel measurable. That is,  $\mathcal{Q} \in \mathcal{B}(\mathcal{R})$ .

#### Answer:

1) We know by De Morgan's law that  $A \cup B = (A^c \cap B^c)^c$ . Let  $A, B \in \Sigma$ :

$$A, B \in \Sigma \implies A^c, B^c \in \Sigma$$
 By property 2  
 $\implies A^c \cap B^c \in \Sigma$  By property 3  
 $\implies (A^c \cap B^c)^c \in \Sigma$  By property 2  
 $\implies A \cup B \in \Sigma$  By De Morgan's

2) Let  $A \in \Sigma$  then  $A^c \in \Sigma$  by property 2. Now,

$$A \cap A^c = \emptyset \in \Sigma$$
 by property 3  $\emptyset^c = \Omega \in \Sigma$  by property 2

Thus,  $\emptyset, \Omega \in \Sigma$ .

3) If  $A \in \Sigma$  then  $A^c \in \Sigma$  necessarily. Also, by 2),  $\emptyset, \Omega \in \Sigma$ , so the smallest possible set is:

$$\Sigma = \{\emptyset, A, A^c, \Omega\}$$

This is clearly closed under complementation. It is also very quick to verify that this set is closed under intersection:

$$X \cap \Omega = X \in \Sigma$$
, for all  $X \in \Sigma$   
 $X \cap \emptyset = \emptyset \in \Sigma$ , for all  $X \in \Sigma$   
Lastly,  $A \cap A^c = \emptyset \in \Sigma$ 

4) Recall that  $\mathcal{B}(\mathbb{R})$  is the  $\sigma$ -algebra formed by the open sets of  $\mathbb{R}$ . Equivalently, it is also formed by the closed sets. Also recall the facts from real analysis that singletons are closed and the rational numbers are countable. Using 1) that  $\mathcal{B}(\mathbb{R})$  is closed under countable union,

$$\mathbb{Q} = \big(\bigcup_{q \text{ rational}} q\big) \in \mathcal{B}(\mathbb{R})$$

That is,  $\mathbb{Q}$  is the union of a countable number of closed sets, therefore,  $\mathbb{Q} \in \mathcal{B}(\mathbb{R})$ 

**Exercise 2:** let P be a probability measure on  $(\Omega, \Sigma)$ . Only utilizing the definition of probability measure given in the class, solve the following problem.

- 1) Show that for any  $A, B \in \Sigma$  satisfying  $A \subseteq B$ , we have  $0 \le P(A) \le P(B)$ .
- 2) Show that for any positive k, we have

$$P\big(\bigcup_{i=1}^k A_i\big) \le \sum_{i=1}^k P(A_i)$$

3) Does the previous inequality still hold when  $k = \infty$ ?

### Answer:

1) In general, we can write  $B = B \setminus A \cup B \cap A$ . But,  $B \cap A = A$  since  $A \subseteq B$ . So,  $B = B \setminus A \cup A$ . By definition,  $B \setminus A$  and A are disjoint. So we can use the property of probability measures:

$$P(B) = P(B \setminus A \cup A) = P(B \setminus A) + P(A)$$
  
 $P(B) > P(A) \text{ since } P(B \setminus A) > 0$ 

Also,  $P(A) \ge 0$  since it is a probability measure. Thus,  $0 \le P(A) \le P(B)$ .

2) We use a similar idea as in 1). We construct a new sequence that is a partition. Define:

$$B_1 = A_1$$

$$B_i = A_i \setminus (A_{i-1} \cup A_{i-2} \cup \ldots \cup A_1)$$

By construction,  $B_i \subseteq A_i$  and  $\bigcup_{i=1}^k A_i = \bigcup_{i=1}^k B_i$ . So,

$$P(\bigcup_{i=1}^{k} A_i) = P(\bigcup_{i=1}^{k} B_i) = \sum_{i=1}^{k} P(B_i)$$

Since  $B_i \subseteq A_i$ , by the previous part,

$$\sum_{i=1}^k P(B_i) \le \sum_{i=1}^k P(A_i)$$

Which gives us the result  $P(\bigcup_{i=1}^k A_i) \leq \sum_{i=1}^k P(A_i)$ .

3) Yes, the inequality still holds if  $k = \infty$ . Let  $\lim_{i \to \infty} A_i = A$ . Then by the continuity of the probability measures,

$$\lim_{i \to \infty} P\left(\bigcup_{i=1}^{k} A_i\right) = P\left(\lim_{i \to \infty} \bigcup_{i=1}^{k} A_i\right) = P(A)$$

Now taking the limit on the right side of the inequality,

$$\lim_{i \to \infty} \sum_{i=1}^{k} P(A_i) = \sum_{i=1}^{\infty} P(A_i)$$

Thus, since the left hand side converges,

$$\lim_{i \to \infty} P\left(\bigcup_{i=1}^{k} A_i\right) \le \lim_{i \to \infty} \sum_{i=1}^{k} P(A_i)$$

$$\implies P\left(\bigcup_{i=1}^{\infty} A_i\right) = P(A) \le \sum_{i=1}^{\infty} P(A_i)$$

**Exercise 3:** For any measurable function f, show that

$$\left| \int f dP \right| < \infty$$
 if and only if  $\int |f| dP < \infty$ 

**Answer:** Suppose that  $\left| \int f dP \right| < \infty$ ,

$$\left| \int f dP \right| = \left| \int f^+ dP - \int f^- dP \right| < \infty$$

Then f is integrable by definition since this implies  $\int f^+ dP < \infty$  and  $\int f^- dP < \infty$ . Note that  $|f| = f^+ + f^-$ . So,

$$\int |f|dP = \int f^+ + f^- dP = \int f^+ dP + \int f^- dP$$

But we have shown that both of these integrals are finite. So,

$$\int |f|dP < \infty$$

Now suppose  $\int |f| dP < \infty$ . Note that,

$$\int f^+ dP \le \int |f| dP < \infty \text{ and}$$
$$\int f^- dP \le \int |f| dP < \infty$$

Which implies  $\left| \int f dP \right| = \left| \int f^+ dP - \int f^- dP \right| < \infty$ .

**Exercise 4:** This exercise consists of two questions, concerns the  $\sigma$ -finiteness of a measure.

- 1) Show that the Lebesgue measure on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  is  $\sigma$ -finite.
- 2) Show that the counting measure on  $(\Omega, 2^{\Omega})$  is  $\sigma$ -finite if and only if  $\Omega$  is countable.

## Answer:

1) Define the sequence of  $A_i$ 's as follows,

$$A_i = (i, i+1]$$

Then we have  $\mu(A_i) = 1$  for  $i \in \mathbb{Z}$  and  $\bigcup_{i \in \mathbb{Z}} A_i = \mathbb{R}$ . Thus the Lebesgue measure on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  is  $\sigma$ -finite.

2) Suppose  $\Omega$  is countable. That is, without loss of generality,  $\Omega = \mathbb{N}$ . Define the sequence of  $A_i = i$ . Then,  $P(A_i) = 1$  for all  $i \in \mathbb{N}$  and  $\bigcup_{i \in \mathbb{N}} A_i = \mathbb{N}$ . Thus, if  $\Omega$  is countable, the counting measure is  $\sigma$ -finite. Now suppose that the counting measure is  $\sigma$ -finite. Thus, there exists  $\{A_i\}$  measurable with  $\bigcup_{i \in \mathbb{N}} A_i = \Omega$  and  $P(A_i) < \infty$  by definition. However, this is a countable union of sets with a finite number of elements. Thus,  $\bigcup_{i \in \mathbb{N}} A_i \neq \mathbb{R}$ . Therefore,  $\Omega$  is countable.

**Exercise 5:** Let  $X : \Omega \to \mathbb{R}$  be a discrete random variable on probability space  $(\Omega, \Sigma, P)$  and denote the corresponding induced measure by  $P_X$ . We define the support of  $P_X$  as

$$\Omega_X = \{ x \in \mathbb{R} : P(X = x) > 0 \}$$

Please answer the following two questions.

- 1) First assume that  $|\Omega_X| < \infty$ , that is,  $\Omega_X$  contains finite numbers of elements. Show that the probability mass function (pmf) of X, denoted f, is indeed the density of  $P_X$  with respect to the counting measure on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ .
- 2) Show the same thing when  $|\Omega_X| = \infty$ .

#### Answer:

1) Let # denote the counting measure. Then  $P_X \ll \#$  since,

$$\#(A) = 0 \implies P_X(A) = P(X^{-1}(A)) = P(X^{-1}(\emptyset)) = 0$$

Also, we proved that the counting measure # is  $\sigma$ -finite if  $\Omega_X$  countable in the previous question. Since it is finite, it is countable. By Radon-Nikodym theorem, if we restrict ourselves to  $\Omega_X$ , then we know a density exists. In this specific case, we do not need  $\sigma$ -finiteness. We only have to show that  $P_X(A) = \int_A f d\#$  to show that f is the density. Since  $\Omega_X$  is finite, then  $f: \Omega \to \mathbb{R}$  is simple and has the form,

$$f = \sum_{x \in A} f(x) 1_{\{x\}}$$

Putting this into the integral above,

$$\int_{A} f d\# = \int_{A} \sum_{x \in A} f(x) 1_{\{x\}} d\# = \sum_{x \in A} f(x)$$

Which is exactly what we desire,

$$P_X(A) = P(X^{-1}(A)) = P(\omega : X(\omega) \in A) = \sum_{x \in A} f(x)$$

Where the last equality is the definition of the pmf Thus,  $P_X(A) = \int_A f d\#$  iff f is the pmf. 2) If  $|\Omega_X| = \infty$  then we must construct our f different. Since  $\Omega_X$  is  $\sigma$ -finite, we can construct an increasing sequence of sets  $(A_i)$  such that  $\bigcup_i A_i = \Omega_X$  and  $A_{i-1} \subset A_i$ . Then define  $f_n$  as,

$$f_n = \sum_{x \in A_n} f(x) \mathbb{1}_{\{x\}}$$

Since  $(f_n)$  is an increasing sequence of simple functions with  $f_n \to f$ . Then,

$$\int_{\Omega_X} f d\# = \lim_{n \to \infty} \int_{\Omega_X} f_n d\# = \lim_{n \to \infty} \int_{\Omega_X} \sum_{x \in A_n} f(x) 1_{\{x\}} d\#$$
$$= \lim_{n \to \infty} \sum_{x \in A_n} f(x) = \sum_{x \in \Omega_X} f(x)$$

We can exchange the limit and the integral due to the monotone convergence theorem. Now, as before,

$$P_X(\Omega_X) = P(X^{-1}(\Omega_X)) = P(\omega : X(\omega) \in \Omega_X) = \sum_{x \in \Omega_X} f(x)$$

Thus,  $P_X(A) = \int_A f d\#$  iff f is the pmf.