

# ORF 524: Statistical Theory and Methods

## Homework 1

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**Exercise 1** (10 points). Recall the definition of  $\sigma$ -Algebra. Let  $(\Omega, \bar{\Sigma})$  be a measurable space, that is,  $\bar{\Sigma}$  satisfies the following three properties:

- $\bar{\Sigma} \neq \emptyset, \bar{\Sigma} \subseteq 2^\Omega$ .
- $A \in \bar{\Sigma}$  implies that  $A^c \in \bar{\Sigma}$ . Here we use  $A^c$  to denote the complement of  $A$ .
- For any  $A_1, A_2, \dots \in \bar{\Sigma}$ , we have  $\bigcap_{i \geq 1} A_i \in \bar{\Sigma}$ .

Based on these properties, solve the following problems.

- (1). Show that  $\bar{\Sigma}$  is closed under union.
- (2). Show that  $\bar{\Sigma}$  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  $\mathbb{Q}$ , is Borel measurable. That is,  $\mathbb{Q} \in \mathcal{B}(\mathbb{R})$ .

*Proof.* We first show that  $\bar{\Sigma}$  is closed under union. Let  $A_1, A_2, \dots \in \bar{\Sigma}$ , by De Morgan's law we have

$$\left( \bigcup_i A_i \right)^c = \bigcap_i A_i^c.$$

Since  $A_i \in \bar{\Sigma}$  for all  $i \geq 1$ , it holds that  $A_i^c \in \bar{\Sigma}$ . Since  $\bar{\Sigma}$  is closed under intersection,  $\bigcap_i A_i^c$  is in  $\bar{\Sigma}$ . Thus we have  $\bigcup_i A_i \in \bar{\Sigma}$ .

For the second problem, we only need to show  $\emptyset \in \bar{\Sigma}$ , because this implies  $\Omega \in \bar{\Sigma}$  by the second property of  $\sigma$ -algebra. Since  $\bar{\Sigma} \neq \emptyset$ , there exists  $A \neq \emptyset$  such that  $A \in \bar{\Sigma}$ . Thus  $A^c$  is also in  $\bar{\Sigma}$ . Since  $\bar{\Sigma}$  is closed under intersection, we have

$$\emptyset = A \cap (A^c) \in \bar{\Sigma},$$

which concludes the proof.

For the third problem, for any  $\sigma$ -algebra  $\bar{\Sigma}$  containing  $A$ ,  $A^c$  is also in  $\bar{\Sigma}$ . Thus we have

$$\bar{\Sigma}_0 := \{\emptyset, \Omega, A, A^c\} \subseteq \bar{\Sigma}.$$

Moreover, it is easy to verify that  $\bar{\Sigma}_0$  itself is a  $\sigma$ -algebra. Thus  $\bar{\Sigma}_0$  is the smallest  $\sigma$ -algebra containing  $A$ . Finally, for the last one, since  $\mathbb{Q}$  is countable, we can enumerate all its elements by

$$\mathbb{Q} = \{a_1, a_2, \dots\}.$$

Thus we can write  $\mathbb{Q} = \bigcup_{i \geq 1} \{a_i\}$ . Since  $\{a_i\}^c = (-\infty, a_i) \cup (a_i, +\infty)$  is a union of two open sets,  $\{a_i\}^c \in \mathcal{B}(\mathbb{R})$ . This implies that  $\{a_i\} \in \mathcal{B}(\mathbb{R})$ . Thus we conclude that  $\mathbb{Q} \in \mathcal{B}(\mathbb{R})$ .  $\square$

**Exercise 2** (10 points). Let  $P$  be a probability measure on  $(\Omega, \bar{\Sigma})$ . Only utilizing the definition of probability measure given in the class, solve the following problems.

(1). Show that for any  $A, B \in \bar{\Sigma}$  satisfying  $A \subseteq B$ , we have  $0 \leq P(A) \leq P(B)$ .

(2). Show that for any positive integer  $k$ , we have

$$P\left(\bigcup_{i=1}^k A_i\right) \leq \sum_{i=1}^k P(A_i). \quad (1)$$

(3). Does inequality (1) still hold when  $k = \infty$ ?

*Proof.* For the first problem, we write  $B = A \cup (B \setminus A)$ . Note that  $A$  and  $B \setminus A$  are disjoint. We set  $A_1 = A$ ,  $A_2 = B \setminus A$ , and  $A_i = \emptyset$  for  $i \geq 3$ , then these sets are disjoint. Since  $P(\emptyset) = 0$ , we have

$$P(B) = P\left(\bigcup_i A_i\right) = \sum_i P(A_i) = P(A) + P(B \setminus A) \geq P(A), \quad (2)$$

where the last inequality follows from the non-negativity of probability measure.

For the second problem, we prove by induction. For  $k = 1$ , the inequality (1) holds trivially. Suppose the argument holds for  $k = m$ . Now we set  $k = m + 1$ . Note that we have

$$\bigcup_{i=1}^{m+1} A_i = A_{m+1} \cup \left[ \bigcup_{i=1}^m (A_i \setminus A_{m+1}) \right]. \quad (3)$$

Similar to (2) we have

$$P\left(\bigcup_{i=1}^{m+1} A_i\right) = P(A_{m+1}) + P\left[\bigcup_{i=1}^m (A_i \setminus A_{m+1})\right]. \quad (4)$$

By the induction assumption, the desired inequality holds for  $k = m$ , we have

$$P\left[\bigcup_{i=1}^m (A_i \setminus A_{m+1})\right] \leq \sum_{i=1}^m P(A_i \setminus A_{m+1}) \leq \sum_{i=1}^m P(A_i), \quad (5)$$

where the last inequality follows from (2). Combining (4) and (5), we have

$$P\left(\bigcup_{i=1}^{m+1} A_i\right) \leq \sum_{i=1}^{m+1} P(A_i).$$

Therefore, the desired inequality holds for any integer  $k > 0$ .

For  $k = \infty$ , the argument still holds. If  $\sum_{i=1}^{\infty} P(A_i) = +\infty$ , the inequality holds trivially. Thus we only need to consider the case where  $\sum_{i=1}^{\infty} P(A_i) < +\infty$ . In this case, note that

$$\bigcup_{i=1}^{\infty} A_i = A_1 \bigcup_{i=1}^{\infty} \left[ A_i \setminus \left( \bigcup_{j=1}^{i-1} A_j \right) \right], \quad (6)$$

where we denote  $A_0 = \emptyset$ . Since the sets  $\{A_i \setminus (\bigcup_{j=1}^{i-1} A_j), i \geq 1\}$  are disjoint, we have

$$P\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} P\left[A_i \setminus \left(\bigcup_{j=1}^{i-1} A_j\right)\right] \leq \sum_{i=1}^{\infty} P(A_i).$$

Thus we conclude the proof.

**Another proof of the third problem:**

As another proof, we show that the continuity of probability measure implies (1) for  $k = \infty$ .

**Theorem 1** (continuity of probability measure). *Let  $(\Omega, \mathcal{B}(\Omega), P)$  be a probability space. Let  $\{B_n\}_{n=1}^{\infty}$  be a sequence of increasing sets such that  $B_n \subseteq B_{n+1}$  for  $n \geq 1$  and  $B = \bigcup_{n=1}^{\infty} B_n$ . Then it holds that*

$$\lim_{n \rightarrow \infty} P(B_n) = P(B). \quad (7)$$

Now we apply this theorem by letting  $B_n = \bigcup_{i=1}^n A_i$  in (7) for all  $i \geq 1$ . Then set  $B$  in Theorem 1 is just  $\bigcup_{i=1}^{\infty} A_i$ . By Theorem 1 we have

$$\lim_{n \rightarrow \infty} P\left(\bigcup_{i=1}^n A_i\right) = P\left(\bigcup_{i=1}^{\infty} A_i\right). \quad (8)$$

By the second problem, for each finite  $n$ , it holds that

$$P\left(\bigcup_{i=1}^n A_i\right) \leq \sum_{i=1}^n P(A_i). \quad (9)$$

Thus letting  $n$  goes to infinity in (9) and combining (8), we finally have

$$P\left(\bigcup_{i=1}^{\infty} A_i\right) = \lim_{n \rightarrow \infty} P\left(\bigcup_{i=1}^n A_i\right) \leq \sum_{i=1}^{\infty} P(A_i),$$

which concludes the proof. □

In what follows we prove Theorem 1.

*Proof of Theorem 1.* For each  $n \geq 1$ , we define  $C_n = B_n \setminus B_{n-1} = B_n \cap B_{n-1}^c$ , where we set  $B_0 = \emptyset$ . Then  $\{C_n\}_{n=1}^{\infty}$  is a sequence of disjoint sets. Moreover, we have  $B_n = \bigcup_{i=1}^n C_i$  and  $B = \bigcup_{i=1}^{\infty} C_i$ . Thus we have

$$\lim_{n \rightarrow \infty} P(B_n) = \lim_{n \rightarrow \infty} \sum_{i=1}^n P(C_i) = \sum_{i=1}^{\infty} P(C_i) = P(B).$$

. Thus we conclude the proof. □

**Exercise 3** (10 points). *Show that*

$$\left| \int f dP \right| < \infty \text{ if and only if } \int |f| dP < \infty.$$

*Proof.* We define  $f_+ = f \cdot \mathbb{1}\{f \geq 0\}$  and  $f_- = -f \cdot \mathbb{1}\{f \leq 0\}$ . By definition, we have

$$\int f dP = \int f_+ dP - \int f_- dP, \quad (10)$$

where at least one of the two terms on the right-hand side of (10) is finite. Thus

$$\left| \int f dP \right| < \infty \text{ is equivalent to } \int f_+ dP < \infty \text{ and } \int f_- dP < \infty.$$

Note that  $f_+$  and  $f_-$  are both non-negative. Thus it follows that

$$\left\{ \int f_+ dP < \infty \text{ and } \int f_- dP < \infty \right\} \text{ is equivalent to } \int (f_+ + f_-) dP < \infty.$$

Since  $|f| = f_+ + f_-$ , we conclude that

$$\left| \int f dP \right| < \infty \text{ and } \int |f| dP < \infty$$

are equivalent. □

**Exercise 4** (10 points). *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.*

*Proof.* To see that the Lebesgue measure is  $\sigma$ -finite, let  $A_i = [i, i+1)$  for  $i \in \mathbb{Z}$ . Then we have  $A_i \in \mathcal{B}(\mathbb{R})$  and

$$\mathbb{R} = \bigcup_{i \in \mathbb{Z}} A_i.$$

Moreover, the Lebesgue measure of  $A_i$  is one for each  $i \in \mathbb{Z}$ . Thus the Lebesgue measure is  $\sigma$ -finite.

Now we consider the second question. Let  $\nu$  be the counting measure on  $(\Omega, 2^\Omega)$ . From the definition of  $\sigma$ -finiteness, if  $\nu$  is  $\sigma$ -finite, there exists  $\{A_i, i \geq 1\} \subseteq 2^\Omega$  such that  $\nu(A_i) < \infty$  and  $\bigcup_{i \geq 1} A_i = \Omega$ . Thus each  $A_i$  contains finite number of elements in  $\Omega$ , which implies that

$$\Omega = \bigcup_{i \geq 1} A_i$$

is the countable union of sets each containing finite number of elements. Thus  $\Omega$  is countable.

On the other hand, if  $\Omega$  is countable, we could list its elements by

$$\Omega = \{a_1, a_2, \dots\}.$$

Let  $A_i = \{a_i\}$ , then we have  $\bigcup_{i \geq 1} A_i = \Omega$  and  $\nu(A_i) = 1 < \infty$ . Thus the counting measure  $\nu$  is  $\sigma$ -finite. □

**Exercise 5** (10 points). Let  $X: \Omega \rightarrow \mathbb{R}$  be a discrete random variable on probability space  $(\Omega, \bar{\Sigma}, P)$  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 number of elements. Show that the probability mass function (pmf) of  $X$ , denoted by  $f$ , 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  $|\Omega_X| = \infty$ .

**Remark.** In this exercise, the counting measure on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  is not  $\sigma$ -finite, but this does not contradict Radon-Nikodym theorem. Recall that Radon-Nikodym theorem states that, if a measure  $\mu$  is absolutely continuous with respect to a  $\sigma$ -finite measure  $\nu$ , then there exists a function  $f: \mathcal{X} \rightarrow \mathbb{R}_+$  such that, for any measurable set  $A$ ,

$$\mu(A) = \int_A f d\nu. \quad (11)$$

However, even if  $\nu$  is not  $\sigma$ -finite, as long as (11) holds for any measurable set  $A$ , we might still call  $f$  the density of  $\mu$  with respect to  $\nu$ .

Therefore, in this exercise, to show that the pmf is the density of  $P_X$  with respect to the counting measure on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ , which is denoted by  $\nu$ , we only need to show that for any  $A \in \mathcal{B}(\mathbb{R})$ , we have

$$P_X(A) = \int_A f d\nu. \quad (12)$$

Since  $X$  is discrete, in other words,  $X$  only takes countable number of values, we have

$$P_X(A) = P_X(A \cap \Omega_X) \text{ and } \int_A f d\nu = \int_{A \cap \Omega_X} f d\nu.$$

Thus, we only need to show (12) for any  $A \in 2^{\Omega_X}$ . In this case, we only need to verify that

$$\sum_{a \in A} f(a) = \int_A f d\nu$$

holds for any  $A \in 2^{\Omega_X}$ .

**Proof. Question (1).** Note that  $\Omega_X$  is countable since  $X$  is discrete. We first assume that  $\Omega_X$  is a finite set. Let  $\nu$  be the counting measure on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ , we show that the pmf of  $X$  is the density of  $P_X$  with respect to  $\nu$ . By the definition of density, we need to show that for any  $A \subset \mathcal{B}(\mathbb{R})$ , it holds that

$$P_X(A) = \int f \cdot \mathbb{1}_A d\nu. \quad (13)$$

Since  $X$  is discrete, we have  $P_X(A) = P_X(A \cap \Omega_X)$ . In addition, for any  $x \in \mathbb{R}$ ,

$$f(x) = \begin{cases} \mathbb{P}(X = x) & x \in \Omega_X, \\ 0 & x \notin \Omega_X, \end{cases}$$

which implies that  $f = f \cdot \mathbb{1}_{\Omega_X}$ . Thus we have

$$\int f \cdot \mathbb{1}_A d\nu = \int f \cdot \mathbb{1}_{\Omega_X} \cdot \mathbb{1}_A d\nu = \int f \cdot \mathbb{1}_{A \cap \Omega_X} d\nu.$$

Therefore, to show (13) for any  $A \in \mathcal{B}(\mathbb{R})$ , it suffices to show that (13) holds for any  $A \in 2^{\Omega_X}$ .

In what follows, for any  $A \in 2^{\Omega_X}$ , we show that (13) holds. When  $|\Omega_X| < \infty$ , we can write  $A = \{a_1, a_2, \dots, a_m\}$  where  $m \leq |\Omega_X|$ . Then we have

$$P_X(A) = P(X \in \{a_1, \dots, a_m\}) = \sum_{i=1}^m f(a_i) = \sum_{a \in \Omega_X} f(a) \cdot \mathbb{1}_A(a) \cdot \nu(a) = \int f \cdot \mathbb{1}_A d\nu.$$

Thus we conclude the proof.

**Question (2).** When  $|\Omega_X| = \infty$ , the result still holds. Let  $\nu$  be the counting measure on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$ . Similar to the last question, we only need to show (13) for any  $A \in 2^{\Omega_X}$ . In what follows, we assume that  $|A| = \infty$ , since the case where  $A$  is finite is already established. Without loss of generality, we can write  $A \in 2^{\Omega_X}$  by

$$A = \{a_1, a_2, \dots\} = \{a_i : i \geq 1\} = \bigcup_{i=1}^{\infty} \{a_i\}.$$

Then by the definition of  $P_X$ , we have

$$P_X(A) = P(X \in A) = \sum_{i=1}^{\infty} f(a_i) \leq 1, \quad (14)$$

where the last inequality holds because  $P_X$  is a probability measure. By the definition of integration, we have

$$\int f \cdot \mathbb{1}_A d\nu = \sup \int g d\nu, \quad (15)$$

where the supremum is taken over all simple function  $g$  such that  $0 \leq g \leq f \cdot \mathbb{1}_A$ . Since  $f \cdot \mathbb{1}_A$  is supported on  $A$  and  $0 \leq g \leq f \cdot \mathbb{1}_A$ ,  $g$  is supported on a countable subset of  $A$  and for any  $i \geq 1$ , we have  $0 \leq g(a_i) \leq f(a_i)$ . Thus it follows that

$$\int g d\nu \leq \sum_{i=1}^{\infty} g(a_i) \leq \sum_{i=1}^{\infty} f(a_i).$$

Then taking supremum and combining (14), we obtain

$$P_X(A) \geq \int f \cdot \mathbb{1}_A d\nu.$$

Moreover, by (14), for any  $\epsilon > 0$ , there exists an integer  $N$  such that  $\sum_{i>N} f(a_i) < \epsilon$ . We define a simple function  $g_N$  by

$$g_N = f \cdot \mathbb{1}_{\{a_1, a_2, \dots, a_N\}}.$$

Then we have

$$\int g_N d\nu = \sum_{i=1}^N f(a_i) \geq P_X(A) - \epsilon. \quad (16)$$

Taking supremum on the left-hand side of (16), we have

$$\int f \cdot \mathbb{1}_A d\nu \geq P_X(A) - \epsilon,$$

where  $\epsilon$  can be set arbitrarily small. Therefore, we conclude that

$$P_X(A) = \int f \cdot \mathbb{1}_A d\nu,$$

which implies that  $f$  is the density of  $P_X$  with respect to  $\nu$ . □