

# ST5215 Advanced Statistical Theory, Lecture 1

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**Pre-Requisite:** ST2131 and ST2132 or Departmental approval

**Modes:** 1.5h lecture + 0.5h tutorial (tentative)

**Online teaching:**

- Disable your webcam to release network demand
- For questions: “raise hands” or text chat in Zoom
- Keep your microphone muted unless speaking

**Office Hours:** Tue/Thur 16:00 - 17:00PM via Zoom (tentative)

**Textbook:**

- Jun Shao, *Mathematical Statistics*, Springer, 2nd Edition, 2003.

**References:**

- L. Wasserman, *All of Statistics*.
- G. Casella and R. L. Berger, *Statistical Inference* 2nd Edition.
- P. J. Bickel and K. A. Doksum, *Mathematical statistics: basic ideas and selected topics*.
- R.W. Keener, *Theoretical Statistics: Topics for a Core Course*.
- E.L. Lehmann G. Casella, *Theory of Point Estimation*.

**Grading Policy:** 4 homework ( $5\% \times 4 = 20\%$ ), 1 midterm (30%) and 1 final (50%).

## Homework:

- Submitted to **LumniNUS/Files/Submissions/HWXX** by 2pm on the due date.
- Late submission will never be accepted
- Homework submission must be typeset ( $\text{\LaTeX}$ , Mathpix ...)
- You can discuss with your classmates on homework but you **must** write up your solutions independently in your own words
- You will get zero for a submission if someone's is identical or very similar to yours.

## Exam:

- Midterm: Thur, 01 Oct 2020, 14:00-16:00
- Final: Wed, 02 Dec 2020, 13:00 - 15:00
- Closed-book (No books, notes, or electronics)
- Online (you will be supervised via Zoom)

## Topics and chapters:

- Probability theory: Ch 1.1 – 1.4.3
- Models and statistics: Ch 2.1 – 2.2
- Point estimators (the method of moments, the method of maximum likelihood)
- Decision theory: Ch 2.3
- UMVUE: Ch 3.1.1 – 3.1.3
- Bayes estimation and risk: Ch 4.1, 4.3
- Asymptotics: Ch 1.5, 2.5, 4.5
- Other topics (Linear model: Ch 3.3; U-statistics: Ch 3.2)

# Measure and $\sigma$ -fields

- Let  $\Omega$  be a set of objects. *outcome space*; *sample space*.
- Want to measure the “size” of subsets of  $\Omega$ 
  - ▶ If  $\Omega = \mathcal{R}$  and  $A = [a, b]$ , then the size of  $A$  is naturally its length  $b - a$ .
  - ▶ If  $\Omega = \mathcal{R}^2$  and  $A$  is a polygon, its size is defined as its area
- For a general  $\Omega$ , we need to specify which subsets are measurable (i.e., define what can be an “event”)

## Definition

A collection  $\mathcal{F}$  of subsets of a set  $\Omega$  is called a  $\sigma$ -field (or  $\sigma$ -algebra) if

- 1  $\emptyset \in \mathcal{F}$ ,
- 2 if  $A \in \mathcal{F}$ , then  $A^c \in \mathcal{F}$ ,
- 3 if  $A_i \in \mathcal{F}$  for  $i = 1, 2, \dots$ , then  $\bigcup A_i \in \mathcal{F}$ .

A pair  $(\Omega, \mathcal{F})$  of a set  $\Omega$  and a  $\sigma$ -field  $\mathcal{F}$  on  $\Omega$  is called a *measurable space*.

- In probability, a  $\sigma$ -field is a collection of events of interest
- There can be many  $\sigma$ -fields on  $\Omega$
- e.g.:  $\Omega = \{1, 2\}$ ,  $\mathcal{F}_1 = \{\emptyset, \Omega\}$ ,  $\mathcal{F}_2 = \{\emptyset, \{1\}, \{2\}, \Omega\}$

Exercise: Suppose  $(\Omega_1, \mathcal{F}_1)$  and  $(\Omega_2, \mathcal{F}_2)$  are two measure spaces, is  $\mathcal{F}_1 \times \mathcal{F}_2 = \{A_1 \times A_2 : A_i \in \mathcal{F}_i\}$  a  $\sigma$ -field on  $\Omega_1 \times \Omega_2$ ? Why?

**Counter example:**

- $\Omega_1 = \{a, b\}$ ,  $\Omega_2 = \{c, d\}$ ,  $\mathcal{F}_i = 2^{\Omega_i}$  (the power set)
- $K = \{a\} \times \{c\}$  and  $L = \{b\} \times \{d\}$ .
- Note that by definition of the Cartesian product,  
 $K = \{a\} \times \{c\} = \{(a, c)\}$ .
- $K, L \in \mathcal{F}_1 \times \mathcal{F}_2$
- But  $K \cup L = \{(a, c), (b, d)\} \notin \mathcal{F}_1 \times \mathcal{F}_2$

# The smallest $\sigma$ -field

- Given a collection  $\mathcal{C}$  of subsets, there exists a  $\sigma$ -field  $\mathcal{F}$  such that  $\mathcal{C} \subset \mathcal{F}$  and if  $\mathcal{E}$  is a  $\sigma$ -field that also contains  $\mathcal{C}$ , then  $\mathcal{F} \subset \mathcal{E}$ . (See Ex 1.6.2)

Such  $\mathcal{F}$  is the *smallest*  $\sigma$ -field that contains  $\mathcal{C}$ . It is denoted by  $\sigma(\mathcal{C})$  and is called the  $\sigma$ -field generated by  $\mathcal{C}$

- Particularly, when  $\Omega = \mathcal{R}$  and  $\mathcal{O}$  is all open sets, we call  $\sigma(\mathcal{O})$  the Borel  $\sigma$ -field, denoted by  $\mathcal{B}$ . Elements in  $\mathcal{B}$  are called Borel sets.
- Similarly, when  $\Omega = \mathcal{R}^d$  and  $\mathcal{O}^d$  is all open sets in  $\mathcal{R}^d$ ,  $\sigma(\mathcal{O}^d)$  is the  $d$ -dimensional Borel  $\sigma$ -field, denoted by  $\mathcal{B}^d$



## Exercise

Show that  $\mathcal{B}$  is also the  $\sigma$ -field generated by all closed subset of  $\mathcal{R}$ .

- Let  $\mathcal{C}$  be the collection of all closed subset of  $\mathcal{R}$
- For any  $B \in \mathcal{O}$ , we have  $B^c \in \mathcal{C}$ , so  $B^c \in \sigma(\mathcal{C})$ , so  $B \in \sigma(\mathcal{C})$ ; i.e.  $\mathcal{O} \subset \sigma(\mathcal{C})$
- So  $\sigma(\mathcal{O}) \subset \sigma(\mathcal{C})$
- For any  $A \in \mathcal{C}$ , we have  $A^c \in \mathcal{O}$ , so  $A^c \in \sigma(\mathcal{O})$ , so  $A \in \sigma(\mathcal{O})$ ; i.e.  $\mathcal{C} \subset \sigma(\mathcal{O})$ ,
- So  $\sigma(\mathcal{C}) \subset \sigma(\mathcal{O})$
- That is,  $\sigma(\mathcal{C}) = \sigma(\mathcal{O})$

# Measure

## Definition

A (positive) measure  $\nu$  on a measurable space  $(\Omega, \mathcal{F})$  is a non-negative function  $\nu : \mathcal{F} \rightarrow \mathcal{R}$  such that

- ① (non-negativity)  $0 \leq \nu(A) \leq \infty$  for all  $A \in \mathcal{F}$ ,
- ② (empty is zero)  $\nu(\emptyset) = 0$ , and
- ③ ( $\sigma$ -additivity):  $\sum_{i=1}^{\infty} \nu(A_i) = \nu(\bigcup_{i=1}^{\infty} A_i)$  if  $A_i \in \mathcal{F}$  for  $i = 1, 2, \dots$  and  $A_1, A_2, \dots$  are disjoint.

- The triplet  $(\Omega, \mathcal{F}, \nu)$  is called a *measure space*.
- If  $\nu(\Omega) = 1$ , it is also called a *probability space*, and the number  $\nu(A)$  is interpreted as the probability of the event  $A$  to happen. And  $\nu$  is often denoted by  $P$  or  $\text{Pr}$ .

# Examples

- Counting measure:  $\nu(A)$  = the number of elements in  $A$ ,  $\forall A \subset \Omega$ .  
(Note:  $\nu(A)$  can be  $\infty$ )
- Lebesgue measure: There exists a unique measure  $m$  on  $(\mathcal{R}, \mathcal{B})$  that satisfies  $m([a, b]) = b - a$  for any  $a < b$ .  
When  $\mathcal{R}$  is mentioned, it is by default endowed with the Borel  $\sigma$ -field and Lebesgue measure unless explicitly mentioned

## $\sigma$ -finite

- A measure  $\nu$  on  $(\Omega, \mathcal{F})$  is said to be  $\sigma$ -finite if there exists a sequence of measurable sets  $A_1, A_2, \dots$  such that  $\bigcup A_i = \Omega$  and  $\nu(A_i) < \infty$  for all  $i$ .
- The Lebesgue measure is  $\sigma$ -finite:  $\mathcal{R} = \bigcup_{i=1}^{\infty} A_i$  with  $A_i = [-i, i]$  and  $m(A_i) = 2i < \infty$ .
- All finite measures are  $\sigma$ -finite.

$\sigma$ -finite is required in some important theorems (Radon-Nikodym, Fubini's), and we focus on  $\sigma$ -finite measures in this course.

## Product measure

How to introduce a measure on a *product space*  $\Omega_1 \times \cdots \times \Omega_d$ , like,  $\mathcal{R}^d = \mathcal{R} \times \cdots \times \mathcal{R}$ ?

- For a product space  $\Omega_1 \times \cdots \times \Omega_d$ , where each  $\Omega_i$  is endowed with a  $\sigma$ -field  $\mathcal{F}_i$ , the  $\sigma$ -field generated by  $\prod_{i=1}^d \mathcal{F}_i = \{A_1 \times \cdots \times A_d : A_i \in \mathcal{F}_i\}$  is called the *product  $\sigma$ -field*.
- One can show: for  $\mathcal{R}^d = \mathcal{R} \times \cdots \times \mathcal{R}$ , the product  $\sigma$ -field is the same as  $\mathcal{B}^d$ .

### Proposition

*Suppose  $(\Omega_i, \mathcal{F}_i, \nu_i)$ ,  $i = 1, 2, \dots, d$ , are measure spaces and  $\nu_1, \dots, \nu_d$  are all  $\sigma$ -finite. There exists a unique  $\sigma$ -finite measure on the product  $\sigma$ -field, denoted by  $\nu_1 \times \cdots \times \nu_d$ , such that*

$$\nu_1 \times \cdots \times \nu_d(A_1 \times \cdots \times A_d) = \prod_{i=1}^d \nu_i(A_i) \quad (1)$$

*for all  $A_i \in \mathcal{F}_i$ .*

### Example (Lebesgue measure on $\mathcal{R}^d$ )

- For  $\mathcal{R}^d$ , we use the Lebesgue measure  $m$  on  $(\mathcal{R}, \mathcal{B})$  to define a unique product measure  $m \times \cdots \times m$ .
- It is called the Lebesgue measure on  $(\mathcal{R}^d, \mathcal{B}^d)$ .
- It is the standard/canonical measure on  $\mathcal{R}^d$ .

Again, without otherwise explicitly mentioned,  $\mathcal{R}^d$  is endowed with  $\mathcal{B}^d$ ) and Lebesgue measure.

## Measurable functions

Recall that  $f : \Omega \rightarrow \Lambda$  is a *continuous function* between two topological spaces  $\Omega$  and  $\Lambda$ , if for every open subset  $A$  in  $\Lambda$ ,  $f^{-1}(A) = \{x \in \Omega : f(x) \in A\}$  is an open set in  $\Omega$ .

### Definition

Let  $(\Omega, \mathcal{F})$  and  $(\Lambda, \mathcal{G})$  be two measurable spaces and  $f : \Omega \rightarrow \Lambda$  a function. The function  $f$  is called a *measurable function* from  $(\Omega, \mathcal{F})$  to  $(\Lambda, \mathcal{G})$  if  $f^{-1}(A) \in \mathcal{F}$  for all  $A \in \mathcal{G}$ .

- If  $f$  is measurable from  $(\Omega, \mathcal{F})$  to  $(\mathcal{R}, \mathcal{B})$ , then  $f$  is said to be *Borel measurable* or is called a *Borel function* on  $(\Omega, \mathcal{F})$ .
- If  $f : \mathcal{R} \mapsto \mathcal{R}$  is continuous, then it is Borel measurable (left for homework)
- If  $f$  is measurable from  $(\Omega, \mathcal{F})$  to  $(\Lambda, \mathcal{G})$ , then  $f^{-1}(\mathcal{G}) = \{f^{-1}(A) : A \in \mathcal{G}\}$  is a sub- $\sigma$ -field of  $\mathcal{F}$  (left for exercise)  
Usually denoted by  $\sigma(f)$  and called the  $\sigma$ -field generated by  $f$ .

## Examples

- The indicator function  $I_A$  for a measurable set  $A$  is a Borel function. Here,

$$I_A(x) = \begin{cases} 1 & x \in A, \\ 0 & x \notin A. \end{cases} \quad (2)$$

- A *simple function* of the form

$$f(\omega) = \sum_{i=1}^k c_i I_{A_i}(\omega), \quad (3)$$

is a Borel function for any real numbers  $c_1, \dots, c_k$  and measurable sets  $A_1, \dots, A_k$ .



## Some results

- **Richness:** Proposition 1.4 in Shao 2003 shows that many operations (addition, multiplication, division, composition, sup, limit) of measurable functions preserve the measurability
- **Approximation by simple functions:** For any non-negative Borel function  $f$ , there exists a sequence of simple functions  $\varphi_n$ 's such that  $0 \leq \varphi_1 \leq \varphi_2 \leq \dots \leq f$  and  $\lim_n \varphi_n(x) = f(x)$  for every  $x \in \Omega$   
Let

$$\varphi_n = \sum_{i=0}^{n2^n-1} \cdot \frac{i}{2^n} I_{\frac{i}{2^n} \leq f < \frac{i+1}{2^n}} + n \cdot I_{f \geq n}, \forall n \in \mathcal{N}$$

Check

- ①  $\varphi_n$ 's are simple functions
- ②  $0 \leq \varphi_1(x) \leq \varphi_2(x) \leq \dots \leq f(x), \forall x \in \Omega$
- ③ If  $f(x) = \infty$ , then  $\varphi_n(x) = n \rightarrow \infty$
- ④ If  $n > f(x)$  then  $f(x) - \varphi_n(x) \leq \frac{1}{2^n}$ , which implies  $\lim_n \varphi_n(x) = f(x)$

# Notations for probability

In probability theory, a measurable function is also called a *random element*, and often denoted by capital letters  $X$ ,  $Y$ ,  $Z$ , ....

If  $X$  is real-valued, then it is called a *random variable* (r.v.); if it is vector-valued, then it is called a *random vector*.

# Integration

We introduce the concept of Lebesgue integral by three steps:

- step 1: define integral of “simple” functions – easy case
- step 2: define integral of non-negative Borel functions by approximation of simple functions
- step 3: define integral of all Borel functions

# Integral of non-negative simple functions

- Suppose  $f : \Omega \rightarrow \mathcal{R}$  is a simple non-negative function:  
 $f(x) = \sum_{i=1}^k c_i 1_{A_i}(x)$  for  $A_i \in \mathcal{F}$  and  $c_i \geq 0$ .
- Define the integral of  $f$  as

$$\int f \, d\nu = \sum_{i=1}^k c_i \nu(A_i)$$

- Well defined even when  $\nu(A_i) = \infty$  for some  $A_i$ :  
 $c \cdot \infty = \infty$  when  $c > 0$  and  $c \cdot \infty = 0$  when  $c = 0$ .
- Note that  $\int f \, d\nu = \infty$  is possible and allowed.

# Integral of a non-negative Borel function

- Any Borel function  $f$  can be approximated by a sequence of simple functions
- Use the integrals of these simple functions as proxy
- Let  $\mathcal{S}_f$  be the collection of all non-negative simple functions such that  $g \leq f$  if  $g \in \mathcal{S}_f$ .
- Define the integral of  $f$  as

$$\int f \, d\nu = \sup\left\{\int g \, d\nu : g \in \mathcal{S}_f\right\}. \quad (4)$$

Exercise: Show that if a sequence of simple functions  $\varphi_n$ 's satisfies  $0 \leq \varphi_1 \leq \varphi_2 \leq \cdots \leq f$  and  $\lim_n \varphi_n = f$ , then

$$\int f \, d\nu = \lim_n \int \varphi_n \, d\nu$$

# Integral of arbitrary Borel functions

- Divide  $f$  into two parts:  $f = f_+ - f_-$ 
  - ▶ Positive part:  $f_+(x) = \max\{f(x), 0\}$
  - ▶ Negative part:  $f_-(x) = -\min\{f(x), 0\} = \max\{-f(x), 0\}$ .  
Note that the negative part is also a nonnegative function
- Define  $\int f \, d\nu$  as

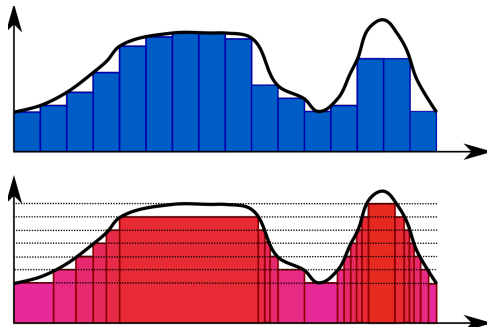
$$\int f \, d\nu = \int f_+ \, d\nu - \int f_- \, d\nu \quad (5)$$

if at least one of  $\int f_+ \, d\nu$  and  $\int f_- \, d\nu$  is finite.

- ▶ if yes, we say the integral of  $f$  *exists*
  - ▶ if not, then we can the integral of  $f$  *does not exist*
- When both  $\int f_+ \, d\nu$  and  $\int f_- \, d\nu$  are finite, we say  $f$  is *integrable*

## Comparison with Riemann Integral

- Both are defined as the limit of a sequence of sums.
- Riemann integral: dividing the  $X$ -domain
- Lebesgue integral: dividing the  $Y$ -range.



- If a function is Riemann integrable on a closed interval, then its Riemann integral will coincide with its Lebesgue integral.
- A bounded function on a closed interval is Riemann integrable if and only if the set of its discontinuities has Lebesgue measure 0.

## Some notations

Integral over a subset  $A \in \mathcal{F}$

- $I_A$  is measurable, and so is the product  $I_A f$ .
- If the integral of  $I_A f$  exists, then we can define

$$\int_A f \, d\nu = \int I_A f \, d\nu. \quad (6)$$

- Notation:  $\int f \, d\nu = \int_{\Omega} f \, d\nu = \int f(x) \, d\nu(x) = \int f(x) \nu(dx)$

For a probability measure  $P$  and a r.v.  $X$ , the expectation of  $X$  is  $\mathbb{E}X = \mathbb{E}(X) = \int X \, dP$



# Exercises

- ① Let  $(\Omega, \mathcal{F})$  be a measurable space and  $C \in \mathcal{F}$ . Show that  $\mathcal{F}_C = \{C \cap A : A \in \mathcal{F}\}$  is a  $\sigma$ -field on  $C$ .
- ② If  $f$  is measurable from  $(\Omega, \mathcal{F})$  to  $(\Lambda, \mathcal{G})$ , then  $f^{-1}(\mathcal{G}) = \{f^{-1}(A) : A \in \mathcal{G}\}$  is a sub- $\sigma$ -field of  $\mathcal{F}$
- ③ Show that if a sequence of simple functions  $\varphi_n$ 's satisfies  $0 \leq \varphi_1 \leq \varphi_2 \leq \dots \leq f$  and  $\lim_n \varphi_n = f$ , then

$$\int f \, d\nu = \lim_n \int \varphi_n \, d\nu$$

# Exercises 1

Let  $(\Omega, \mathcal{F})$  be a measurable space and  $C \in \mathcal{F}$ . Show that  $\mathcal{F}_C = \{C \cap A : A \in \mathcal{F}\}$  is a  $\sigma$ -field on  $C$ .

Proof:

- Since  $\emptyset \in \mathcal{F}$ , so  $\emptyset = C \cap \emptyset \in \mathcal{F}_C$
- If  $C \cap A \in \mathcal{F}_C$ , where  $A \in \mathcal{F}$ .  
Then  $A^c \in \mathcal{F}$ .  
Note that  $(C \cap A)^c$  w.r.t.  $C$  is  $C \setminus (C \cap A) = C \cap A^c \in \mathcal{F}_C$ .
- Suppose  $C \cap A_i \in \mathcal{F}_C$ , where  $A_i \in \mathcal{F}, \forall i \in \mathcal{N}$ .  
 $\cup_i A_i \in \mathcal{F}$   
 $\cup_i (C \cap A_i) = C \cap (\cup_i A_i) \in \mathcal{F}_C$

## Exercises 2

If  $f$  is measurable from  $(\Omega, \mathcal{F})$  to  $(\Lambda, \mathcal{G})$ , then  $f^{-1}(\mathcal{G}) = \{f^{-1}(A) : A \in \mathcal{G}\}$  is a sub- $\sigma$ -field of  $\mathcal{F}$

Proof:

- $\emptyset = f^{-1}(\emptyset) \in f^{-1}(\mathcal{G})$
- If  $f^{-1}(A) \in f^{-1}(\mathcal{G})$ , where  $A \in \mathcal{G}$ ,  
then  $A^c \in \mathcal{G}$ .  
Note that  $(f^{-1}(A))^c = f^{-1}(A^c) \in f^{-1}(\mathcal{G})$ .
- If  $f^{-1}(A_i) \in f^{-1}(\mathcal{G})$ , where  $A_i \in \mathcal{G}$   
then  $\cup_i A_i \in \mathcal{G}$   
Note that  $\cup_i f^{-1}(A_i) = f^{-1}(\cup_i A_i) \in f^{-1}(\mathcal{G})$ .

## Exercises 3

Show that if a sequence of simple functions  $\varphi_n$ 's satisfies  $0 \leq \varphi_1 \leq \varphi_2 \leq \cdots \leq f$  and  $\lim_n \varphi_n = f$ , then

$$\int f \, d\nu = \lim_n \int \varphi_n \, d\nu$$

Proof: “ $\geq$ ” : by definition of  $\int f \, d\nu$

“ $\leq$ ” : If we can prove this result:

### Lemma

For any simple function  $g \leq f$ , we have

$$\int g \, d\nu \leq \lim_n \int \varphi_n \, d\nu.$$

then  $\sup_{g \in S_f} \int g \, d\nu \leq \lim_n \int \varphi_n \, d\nu$ .

For any simple function  $g \leq f$ , we have

$$\int g \, d\nu \leq \lim_n \int \varphi_n \, d\nu.$$

Proof: Suppose  $g = \sum_{j=1}^m b_j I_{B_j}$

- For any  $c \in (0, 1)$ , let  $A_n(c) = \{\varphi_n \geq c \cdot g\}$
- We have  $\varphi_n I_{A_n(c)} \geq c \cdot g I_{A_n(c)}$ .
- Hence

$$\begin{aligned} \int \varphi_n \, d\nu &\geq \int \varphi_n I_{A_n(c)} \, d\nu \geq c \cdot \int g I_{A_n(c)} \, d\nu \\ &= c \sum_j b_j \nu(B_j \cap A_n(c)). \end{aligned}$$

- Since  $\varphi_n \uparrow f$ , we have  $A_n(c) \uparrow \Omega$  (see next slide)
- So  $\nu(B_j \cap A_n(c)) \uparrow \nu(B_j)$  (by Proposition 1.1 in JS)
- Hence

$$\lim_n \int \varphi_n \, d\nu \geq c \sum_j b_j \nu(B_j).$$

Proving  $A_n(c) \uparrow \Omega$ :

Fix any  $x \in \Omega$ .

If  $f(x) = 0$ , then  $\phi_n(x) = g(x) = 0$ , so  $x \in A_n(c)$  for any  $n$  and  $c$ .

If  $f(x) > 0$ , then  $f(x) = \lim \phi_n(x) > c \cdot g(x)$ . So for all  $n$  large enough,  $\phi_n(x) > c \cdot g(x)$ , and  $x \in A_n(c)$ .

This shows  $\lim_n A_n(c) = \cup_n A_n(c) = \Omega$ .

