

- Lecture 1 Introduction: What is real analysis?
 Lecture 2 Elementary measure, Jordan measure
 Lecture 3 Lebesgue outer measure
 Lecture 4 Lebesgue measure
 Lecture 5 Lebesgue measure (continued)
 Lecture 6 Measurable functions
 Lecture 7 Convergences of measurable functions
 Lecture 8 Approximation by simple and/or continuous functions
 Lecture 9 Lebesgue Integrals of non-negative measurable functions
 Lecture 10 Convergence of Lebesgue integrals: Non-negative measurable functions
 Lecture 11 Lebesgue integrals of general measurable functions
 Lecture 12 Convergence of Lebesgue integrals: absolutely integrable functions
 Lecture 13 Lebesgue integrals V.S Riemann integrals, Fubini's theorem

Lecture 14 More on integrals (Lebesgue): Tonelli's theorem, Convolution

- Lecture 15 Abstract measures.
 Lecture 16 Integrations on abstract measure spaces
 Lecture 17 From outer measures to measures
 Lecture 18 Metric v.s. measure
- Lecture 19 The Riesz representation theorem for $C(X)$.
 Lecture 20 L^p Spaces
- Lecture 21 L^p Spaces as metric spaces
 Lecture 22 L^2 Space as inner product space
- Lecture 23 Riesz representation theorem for L^p spaces
 Lecture 24 Signed measure, the Lebesgue-Radon-Nikodym theorem
- Lecture 25 The Lebesgue differentiation theorem
 Lecture 26 The fundamental theorem of calculus
- Lecture 27 The Rademacher differentiation theorem
- For Abstract Measure
- L^p -spaces
- Riesz rep
- differential

答疑. 周日 15:55~17:30

JUR STORY BEGINS

Real Analysis

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- △ Monday 3-5
- △ Thursday 3-4. 5 (答疑)
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Course Information

- △ website: <http://staff.ustc.edu.cn/~wangzuog/course/18S-RealAnalysis/index.html>
(course notes. psets → after each lecture)

△ Reference Books:

- ① 国民强《实变函数论》. ② E. Stein Real Analysis - measure theory, integration & Hilbert Spaces
- ③ T. Tao An introduction to measure theory. Real Analysis

- △ Homeworks: Twice every week after class
Collect on Mondays before class

- △ Score: 30% HW + 30% Midterm + 40% final
A+ essays

- △ Cover:
 - Measure theory
 - Integration theory
 - L^p -Space

Lee 1

What is Real analysis?

Why Should we learn it?

- Wikipedia: Real Analysis deals with real numbers and real-valued functions of real variables. In particular, it deals with the analytic properties of real functions and sequences.

analytic

- limit and convergence
- continuity
- differentiation
- integration

History (Rough)

→ 17th century → 18th century → 19th century

(Newton-Leibniz)
invent Calculus
微积分

(Euler, Laplace...)
Applications of calculus

(Cauchy, Weierstrass...)
rigorous background
(ϵ - δ language)

→ 20th century

(Lebesgue, Banach...)

abstraction → for more applications

Real Analysis (the 20th century's R-A)

Functional Analysis 泛函分析

What did we learn in Mathematical Analysis?

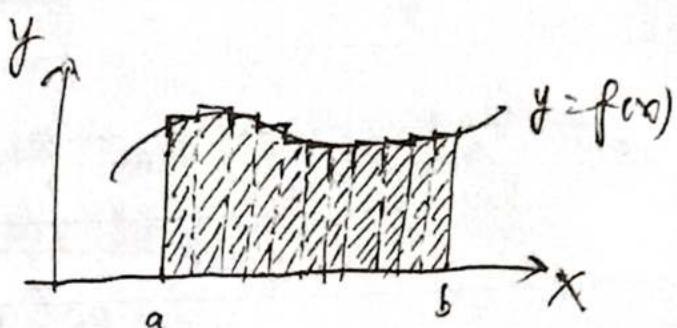
- Sequence and Series of $\begin{cases} \text{real numbers} \\ \text{real functions} \end{cases}$ of complement of \mathbb{R}
 - Taylor series
 - Fourier series

Differentiation $f'(a) = \lim_{h \rightarrow 0} \frac{f(a+h) - f(a)}{h}$ (slope, rate of change)

Riemann Integral

$$\int_a^b f(x) dx = \lim_{n \rightarrow \infty} \sum_i f(t_i) \Delta x$$

Sum up = length of vertical intervals



Relationship between D and I.

The fundamental theorem of calculus

Suppose f is differentiable on $[a, b]$ and f' is Riemann integrable on $[a, b]$. Then $\int_a^b f'(x) dx = f(b) - f(a)$.

Lebesgue theorem // A bounded function f on $[a, b]$ is Riemann integrable if and only if the set of discontinuous points of f is of measure zero

// A set $E \subset \mathbb{R}$ is of measure zero if

$\forall \epsilon > 0 \exists$ intervals (a_i, b_i) s.t.

// $E \subset \bigcup_i (a_i, b_i) \Rightarrow \sum_i (b_i - a_i) < \epsilon$

Shortcome of the Riemann Integration

(i) There exists simple and natural functions which are NOT Riemann integrable

$$\text{e.g. } \chi_\alpha(x) = \begin{cases} 1 & x \in \mathbb{Q} \\ 0 & x \in \mathbb{R} \setminus \alpha \end{cases}$$

- It's discontinuous everywhere \rightarrow Not Riemann integrable

(2) The fundamental theorem of calculus fails if f is not differentiable everywhere or if f' is not Riemann integrable.

e.g. Volterra 1881. $\exists f$ which is differentiable every,
But f' is not Riemann integrable.

(3) $f_n \rightarrow f$. f_n continuous $\nRightarrow f$ continuous, or $\int_a^b f_n(x)dx \rightarrow \int_a^b f(x)dx$
 Need: uniformly convergence
 e.g. \exists $\delta \in \mathbb{R}$ s.t. continuous $f_n \nrightarrow f$ as $n \rightarrow \infty$. But: f is not Riemann integrable

(4) Besicovitch's example: $\exists f$
 defined on the plane. Riemann integrable.
 But, No matter how you choose orthogonal axis x and y .
 $\exists y$ s.t: $f(x,y)$ (as a func of x) is not Riemann integrable
 In particular. $\underset{\text{double integral}}{\iint f dA} \neq \underset{\text{iterated integral}}{\int (\int f(x,y) dx) dy}$
 iterated integral does not exist.

- closed related to Kakeya needle problem

Find a set in \mathbb{R}^n as small as possible. s.t:
 it contains unit length intervals of each direction.

(5) In modern analysis, one regardless of functions is as "points" in some "Space of functions".
 one can define "distance" between two functions
 c.g. $[a,b]$. $d(f,g) = \int_a^b |f(x) - g(x)| dx$.

For example $R =$ all Riemann Integrable functions on $[a,b]$
 \leadsto "metric Space" (R, d)

~ Cauchy Sequence in (R, d) FACT. (R, d) is NOT complete
 ↓
 which in Functional Analysis is crucial

(6) Fourier Series. Suppose f is Riemann integrable on $[-\pi, \pi]$,
 then f represented by $\sum_{n=-\infty}^{\infty} a_n e^{inx}$, $a_n = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(x) e^{-inx} dx$.

We Know:

$$\sum_{n=-\infty}^{\infty} |a_n|^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} |f(x)|^2 dx \quad (\text{Parseval Identity}).$$

Question:

Given a sequence (a_n) can we find a function of s.t.: its Fourier coefficients are a_n s?

In particular, if a_n only satisfies, $\sum |a_n|^2 < \infty$.

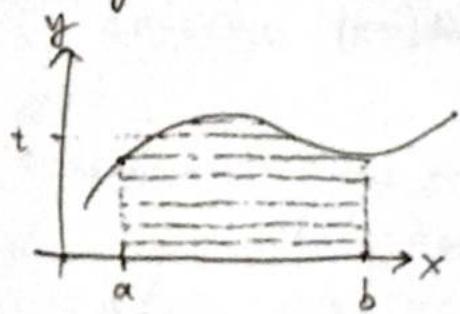
can we find such as f ?

$$\sum a_n e^{inx} \quad \text{if } \sum |a_n| < \infty \rightarrow \text{convergence.}$$


 add these "functions" into the space of consideration

Conclusion: Need new theory of integration

Rough idea & advantage of Lebesgue theory



$$\int_a^b f(x) dx \stackrel{f \text{ is nonnegative}}{=} \int_0^\infty m(\{x : f(x) > t\}) dt.$$

measure = size of set

- (1) χ_Q is Lebesgue integrable and $\int_a^b \chi_Q(x) dx = 0$.
- (2) The fundamental theorem of calculus holds for a very wide class of functions (Absolutely continuous).
- (3) Simple criteria for convergence.
"Dominated convergence theorem"
- (4) "Fubini theory": Double integral = Iterated integral.
平行积分。累次积分。
- (5) $L^p([a,b]) = \{f : f' \text{ is Lebesgue integrable}\}$. | complete
 $d(f,g) = \left(\int_a^b |f-g|^p dx \right)^{\frac{1}{p}}$ ($p \geq 1$)
- (6) Plancherel's theorem: $\sum |a_n|^2 < \infty \rightarrow \exists! f$ Lebesgue integrable whose Fourier coeff are a_n

Moreover:

The process of establishing Lebesgue theory shed a light on how to define measure/integration theory in abstract sets

Key conceptions in Lebesgue's theory:

- measure = "size of a set"
interval \rightarrow length rectangle \rightarrow area solid \rightarrow volume
- we mainly concentrate on subsets E of \mathbb{R} .

A measure m "should" be a function
 $m : P(\mathbb{R}) \xrightarrow{\downarrow} [0, +\infty] \cup \{+\infty\}$.
 all subsets of \mathbb{R} .

such that

$$(1) m([a, b]) = b - a.$$

$$(2) m(\underline{E+x}) = m(E)$$

平行不變性

$$(3) \text{ Suppose } E_i \in P(\mathbb{R}), E_i \cap E_j = \emptyset \ (i \neq j). \\ \text{then } m(\bigvee_i E_i) = \sum_i m(E_i)$$

Unfortunately, such a "measure" does not exist if we allow axiom of choice.

Reason: On $[0, 1]$, we define $x \sim y \Leftrightarrow x - y \in \mathbb{Q}$.

• Let $E = \{ \text{one number from each equivalence class} \} \subset [0, 1]$

• For any $r \in \mathbb{Q}$, let

$$E_r = \{ x+r \pmod{1} \mid x \in E \} \subset [0, 1].$$

$$\text{Then } [0, 1] = \bigcup_{r \in \mathbb{Q}} E_r. \quad m(E_r) = m(E).$$

$$\Rightarrow 1 = m([0, 1]) = m(\bigvee E_r) = \sum m(E_r) = \sum m(E) = \infty \cdot m(E)$$

$$\Rightarrow m(E) = 0.$$

$$\Rightarrow m([0, 1]) = \sum m(E_r) = 0. \quad (\text{矛盾})$$

Even if we replace (3) by (3)': $m(\bigvee_{i=1}^N E_i) = \sum_{i=1}^N m(E_i)$,
 the measure still doesn't exist. (Banach-Tarski paradox)

→ One can't define measure on all subsets of \mathbb{R}

→ measurable sets

→ measurable functions \rightarrow f.o.r a.e. almost everyt. the set.

$\{x \mid f(x) > t\}$ is measurable set

→ Lebesgue integrable function

A warning: There exists many weird examples in Real Analysis which tell us the boundary of the theory.

REAL ANALYSIS IS NOT A SUBJECT OF WEIRD EXAMPLES, INSTEAD, IT IS A SUBJECT THAT TELLS YOU ALL THESE PROPERTIES STILL HOLDS



Littlewood Three Principles:

- (1) Every measurable set $\overset{\text{(in } \mathbb{R})}{\sim}$ is nearly a finite union of intervals
- (2) Every measurable function is nearly continuous
- (3) Every convergent sequence of measurable function is nearly uniformly convergence

Lec.2

Last time.

- Need a new theory of integration
- Lebesgue's idea: need to measure "complicated sets"
- No ways to define a measure for all sets

Today Elementary }
 Jordan } measure

I Elementary measure:

- $I = [a, b], [a, b), (a, b], (a, b)$ intervals

$$\leadsto m(I) = b - a = |I|$$

(we allow $a=b \leftarrow \begin{cases} [a, a] = \{a\}, \\ (a, a) = \emptyset \end{cases}$)

- $B = I_1 \times I_2 \times \dots \times I_d \leftarrow$ box of dimension d.

$$\leadsto m(B) = |B| = |I_1| \times \dots \times |I_d|$$

Def: A set $E \subset \mathbb{R}^d$ is called an Elementary set:

If it is the union of finitely many boxes

i.e. $E = B_1 \cup \dots \cup B_n$. (Note: These boxes may intersect)

Notation: \mathcal{E}_e = the set of all elementary sets in \mathbb{R}^d .

Observation: If $E, F \in \mathcal{E}_e$ then

$$\cdot E \cup F \in \mathcal{E}_e$$

$$\cdot E \cap F \in \mathcal{E}_e$$

$$\cdot E \setminus F \in \mathcal{E}_e$$

$$\cdot E \triangle F = (E \setminus F) \cup (F \setminus E) = (E \cup F) - (E \cap F) \in \mathcal{E}_e$$

(Boolean Structure)

Topological

Prop: Any elementary set can be represented as the union of finitely many non-intersecting boxes.
 i.e. $E = B_1 \cup B_2 \cup \dots \cup B_n$. ($B_i \cap B_j = \emptyset$ if $i \neq j$)

Proof: First suppose $d=1$: $E = I_1 \cup I_2 \cup \dots \cup I_n$.

Denote the endpoints by a_j, b_j of I_j .

Rearrange these endpoints in increasing order as c_1, c_2, \dots, c_m .

(let $I'_1 = [c_1, c_1]$, $I'_2 = [c_1, c_2]$, $I'_3 = [c_2, c_3]$, \dots , $I'_{m-1} = [c_{m-1}, c_m]$)

(let $K = \{k : I'_k \subset E\}$). Then $E = \bigcup_{k \in K} I'_k$ is a non-intersecting union.

For $d > 1$. Suppose $E = \overbrace{B_1 \cup B_2 \cup \dots \cup B_n}$ (may overlap).

where $B_j = I_{j,1} \times \dots \times I_{j,d}$.

For each l apply the 1-dim argument to $I_{1,l}, \dots, I_{n,l}$.

Then consider all possible Cartesian products

of these new non-intersecting intervals +EVN

→ non-overlapping boxes $\rightarrow E = \bigcup_j B_j$

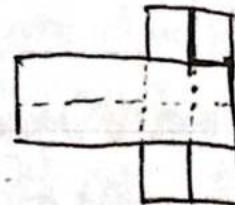
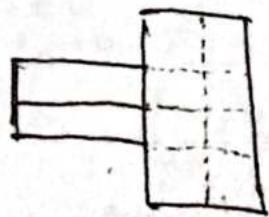
Def: The measure of an elementary set E is

$$m(E) = \sum_{j=1}^n m(B_j)$$

$E = \bigcup B_j$ (maybe overlapped)

One need to check: For two different partitions of E
 into non-overlapping boxes, the formula give the same results

Method 1:



Seek for a common
discretization

Method 2: $m(E) = \lim_{N \rightarrow \infty} \frac{1}{N^d} (\#(E \cap \frac{1}{N} \mathbb{Z}^d))$

"discretization formula": ∫ No partitions here.
 number of only related to E .
 elements in infinitely many sets

The following properties are easy to check via the definition

Prop: The measure of $m: \mathcal{E} \rightarrow [0, +\infty)$ satisfies:

$$(1) m(B) = |B|$$

finite additivity

- \rightarrow (2) If $E_1, E_2 \in \mathcal{E}, E_1 \cap E_2 = \emptyset$ then $(E_1 \cup E_2) = m(E_1) + m(E_2)$
- \nearrow from negative
- (3) If $E \in \mathcal{E}, x \in \mathbb{R}^d$ then $m(E+x) = m(E)$ ← translation-invariance
- (4) If $E_1, E_2 \subset \mathcal{E}$, then $m(E_1 \cup E_2) \leq m(E_1) + m(E_2)$ ← subadditivity
- (5) If $E_1, E_2 \subset \mathcal{E}, E_1 \subset E_2$, then $m(E_1) \leq m(E_2)$ ← monotonicity

Remark One can prove

If a function $\tilde{m}: \mathcal{E} \rightarrow [0, +\infty)$

satisfied (2), (3), with $\tilde{m}([0,1]^d) = 1$.

Then $\tilde{m} = m$. & more radical.

2. Jordan measure

problem of E.M.

triangle, disc are not elementary sets

Archimede's Observation: one can use finitely many boxes : approximate "them both from inside and from outside."

Def: Let $A \subset \mathbb{R}^d$ be any bounded subset

(1) The Jordan inner measure $J_*(A) = \sup_{\substack{\text{fin. many} \\ \mathcal{E} \ni E \subset A}} m(E)$.

exists by (3) mono-

(2) The Jordan outer measure

$J^*(A) = \inf_{\substack{\text{fin. many} \\ \mathcal{F} \ni F \supset A}} m(F) \rightarrow$ exists by non-negative

using finitely many boxes to approximate

(3) If $J^*(A) = J_*(A)$ then we say A is Jordan measurable, with Jordan measure $J(A) = J^*(A) = J_*(A)$.

Notation: \mathcal{T} = the set of all Jordan measurable sets in \mathbb{R}^d .

Remark:

(1) J_* , J^* are defined for any bounded set in \mathbb{R}^d .

(2) J is only defined for "some sets".

(3) $J_* \leq J^*$.

(4) If $A \in \mathcal{E}_e$ then $A \in \mathcal{T}$ and $J(A) = m(A)$.

Example: $A = \mathbb{Q} \cap [0,1]$

$J_*(A) = 0 \neq J^*(A) = 1$. E are formed by finite boxes.

leave out finitely many points in $\mathbb{R} \setminus \mathbb{Q}$.

In fact, $A \in \mathcal{T} \Rightarrow A$ is "nearly" elementary

Thm // Let $A \subset \mathbb{R}^d$ be a bounded subset, Then T.F.A.E.
The followings are equivalent

(1) $A \in \mathcal{T}$

(2) $\forall \varepsilon > 0 \exists E \subset \mathcal{E}_e$ s.t. $E \subset A$ and $m(F \setminus E) < \varepsilon$.

(3) $\forall \varepsilon > 0 \exists E \subset \mathcal{E}_e$ s.t. $E \subset A$ and $J^*(A \setminus E) < \varepsilon$.

$\Leftrightarrow \exists F \subset \mathcal{E}_e$ s.t. $F \supset A$ and $J^*(F \setminus A) < \varepsilon$

a. description of "near" using J^* , well defined

Lemma 1 | Suppose $A_1 \subset A_2$ are bounded
then $J^*(A_1) \leq J^*(A_2)$. Monotonicity
 $J^*(A_1) \leq J_*(A_2)$. In outer and inner measure

Prof: Take a sequences of elementary sets. $E_n \subset A_1 \subset A_2 \subset F_n$

$$\text{St: } m(E_n) \rightarrow J^*(A_1)$$

$$m(F_n) \rightarrow J^*(A_2).$$

$$\text{Then } E_n \subset A_2 \Rightarrow J_*(A_2) = \sup_{E' \subset A_2} m(E') \geq \lim_{n \rightarrow \infty} m(E_n) = J^*(A_1).$$

$$F_n \supset A_1 \Rightarrow J^*(A_1) = \inf_{F' \supset A_1} m(F') \leq \lim_{n \rightarrow \infty} m(F_n) = J^*(A_2).$$

Lemma 2 Suppose A_1, A_2 are bounded finite subadditivity
 Then $J^*(A_1 \cup A_2) \leq J^*(A_1) + J^*(A_2)$. for outer Jordan measure
 * (This property is not available in J^*).

Proof. $A_1 \subset E_n \quad A_1 \cup A_2 \subset \overline{E_n \cup F_n}$ \uparrow Sub-additivity available
 $A_2 \subset F_n$

Lemma 3 Suppose $A \subset B \cap \text{box}$
 Then $A \in \mathcal{T} \Leftrightarrow B \setminus A \in \mathcal{T}$

Proof: $\begin{cases} J^*(B \setminus A) = m(B) - J^*(A) \\ J^*(B \setminus A) = m(B) - J^*(A) \end{cases}$

Proof of Thm.

(1) \Rightarrow (2) Suppose $A \in \mathcal{T}$

Then $\forall \varepsilon > 0 \exists E, F \in \mathcal{L}_n$ s.t.

$$m(F) - \frac{\varepsilon}{2} \leq J^*(A) = J(A) \leq J^*(A) \leq m(E) + \frac{\varepsilon}{2}.$$

$$\Rightarrow m(F) - m(E) \leq \varepsilon.$$

Notice that: $F \supset E$, $F = E \sqcup (F \setminus E)$.

$$\Rightarrow m(F \setminus E) \leq \varepsilon.$$

(2) \Rightarrow (3) Suppose (2) holds. Then $A|E \subset F|E$
 $\Rightarrow J^*(A|E) \leq J^*(F|E) = m(F|E) < \varepsilon.$
 elementary set

(2) \Rightarrow (4) the same as 2) \Rightarrow (3)

(3) \Rightarrow (1) Suppose (3) holds

$\forall \varepsilon > 0 \exists E_\varepsilon \subset A$ s.t. $J^*(A|E_\varepsilon) < \varepsilon$.

Note $E_\varepsilon \in \mathcal{L}_\varepsilon \Rightarrow m(E_\varepsilon) \leq J^*(A)$.

By Lemma 2.

$J^*(A) \leq J^*(A) \leq J^*(A|E_\varepsilon) + J^*(E_\varepsilon) < \varepsilon + m(E_\varepsilon) \leq \varepsilon + J^*(A)$.

let $\varepsilon \rightarrow 0$. $J^*(A) = J^*(A) \Rightarrow A \in \mathcal{T}$

(4) \Rightarrow (1) $\forall \varepsilon > 0 \exists F_\varepsilon \supset A$ s.t. $J^*(F_\varepsilon \setminus A) < \varepsilon$.

W.L.O.G. we can take a box $B \supset F_\varepsilon$ s.t. $B \supset F_\varepsilon$.

let $E_\varepsilon = B \setminus F_\varepsilon \subset B \setminus A$. Then

$J^*(B \setminus A)|E_\varepsilon) = J^*(F_\varepsilon \setminus A) < \varepsilon$.

By (3) \Rightarrow (1), see $B \setminus A \in \mathcal{T}$

By Lemma 3 $A \in \mathcal{T}$.

Remark:

(1) $J: \mathcal{T} \rightarrow [0, +\infty]$, satisfies

* finite additivity

* translation-invariance } with Standardization $J([0,1]) = 1$

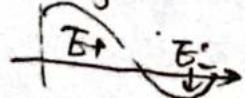
* Subadditivity

* monotonicity

\Downarrow radical properties,

(2) Jordan measure \leftrightarrow Riemann integral

$A \in \mathcal{T} \Leftrightarrow \chi_A$ is Riemann integrable.



$$\int f dx = J(E^+) - J(E^-)$$

Lee. 3

• Shortcome of Jordan measure:

- $\mathbb{Q} \cap [0,1]$ etc. are NOT Jordan measurable

- Not defined for unbounded Sets

→ One way: define the Jordan (outer) measure of any unbounded set to be $+\infty$.But not reasonable to set $J^*(\mathbb{Z}) = +\infty$.- There exists open sets/compact sets which are NOT measurable in Jordan
(¹¹² 例).e.g. $\mathbb{Q} \cap [0,1] = \{q_1, q_2, q_3, \dots\}$.For an $\varepsilon > 0$. (small). We define $\mathcal{Q}_\varepsilon = (q_1 - \frac{\varepsilon}{2}, q_1 + \frac{\varepsilon}{2}) \cup (q_2 - \frac{\varepsilon}{4}, q_2 + \frac{\varepsilon}{4}) \cup \dots \cup (q_n - \frac{\varepsilon}{2^n}, q_n + \frac{\varepsilon}{2^n}) \cup \dots$.Then \mathcal{Q}_ε is an open set, but not Jordan measurable.

$$J^*(\mathcal{Q}_\varepsilon) \geq 1. J^*(\mathcal{Q}_\varepsilon) \leq 2\varepsilon.$$

For compact set:

 $[1,2] \setminus \mathcal{Q}_\varepsilon \leftarrow$ bounded and closed

but not Jordan measurable.

- The "limit" of Jordan measurable sets may be Jordan-unmeasurable e.g. $\mathcal{Q}_N = \{q_1, \dots, q_N\} \xrightarrow[N \rightarrow \infty]{} \mathbb{Q} \cap [0,1]$.

Today: Lebesgue outer measure

$$J^*(A) = \inf_{\substack{\text{E} \ni F \\ E \ni F \supset A}} m(F) = \inf_{\substack{\text{E} \ni F \\ E \ni F \supset A}} \left(\sum_{i=1}^n |B_i| \right) \quad \text{if } A \subset \bigcup_{i=1}^n B_i.$$

Lebesgue's idea:

- Allow the outer measure to take value $+\infty$.
(But not for all unbounded sets). to make the definition make sense
- Allow the number of boxes covering A to be infinitely many.

Def: The Lebesgue outer measure of ANY subset $A \subset \mathbb{R}^d$ is

$$m^*(A) = \inf \left\{ \sum_{i=1}^{\infty} |B_i| \mid B_i \text{ are boxes, } A \subset \bigcup_{i=1}^{\infty} B_i \right\}$$

using countably many boxes. $\boxed{L\text{-cover}}$ I_i 's are open boxes.

Example: $A = Q \cap [0, 1] = \{q_1, q_2, q_3, \dots\}$

Then $A \subset \bigcup_{i=1}^{\infty} [q_i, q_i]$

$$m^*(A) \leq \sum_{i=1}^{\infty} |[q_i, q_i]| = 0 \Rightarrow m^*(A) = 0.$$

For $A \subset Q_\varepsilon$. $m^*(A) \leq \sum_{i=1}^{\infty} 2^{-i}\varepsilon = 2\varepsilon$. $\xrightarrow{\varepsilon \rightarrow 0} m^*(A) = 0$.

Example, Same Argument

\downarrow
Any countable set has Lebesgue outer measure Zero.
e.g. \mathbb{Z}, \emptyset .

Also by definition, one easily see

$$m^*(A) \leq J^*(A).$$

So we get

$$m^*: \mathcal{P}(\mathbb{R}^d) \rightarrow [0, +\infty]$$

闭集

It satisfies the following properties.

we will call them "the axioms of abstract outer measures"

Prop: The Lebesgue outer measure

$$m^*: \mathcal{P}(\mathbb{R}^d) \rightarrow [0, +\infty]$$

Satisfies:

1) (Empty set) $m^*(\emptyset) = 0$.

2) (monotonicity) If $A_1 \subset A_2$, then $m^*(A_1) \leq m^*(A_2)$

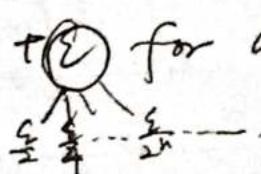
3) (Countable subadditivity)

additivity is only for a measure, not for an outer measure

For any countable many sets A_1, A_2, \dots

$$\text{we have } m^*\left(\bigcup_{n=1}^{\infty} A_n\right) \leq \sum_{n=1}^{\infty} m^*(A_n).$$

Proof:

- (1) $\phi \subset$ any box, $\forall \varepsilon. m^*(\phi) \leq \varepsilon$
- (2) Any coverings of A_2 is also a covering of A_1 .
- (3) (given yourself an ε of room).
try to prove $m^*(\bigcup_{i=1}^{\infty} A_i) \leq \sum_{i=1}^{\infty} m^*(A_i) + \varepsilon$ for any ε .


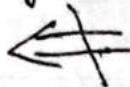
For each A_i , we cover it by $B_{i,1}, B_{i,2}, \dots$

$$A_i \subset \bigcup_{j=1}^{\infty} B_{i,j} \text{ and } \sum_{j=1}^{\infty} |B_{i,j}| \leq m^*(A_i) + 2^{-i} \cdot \varepsilon$$

Consider all boxes $\{B_{i,j}, i=1,2,\dots, j=1,2,\dots\}$.

$$\text{Then } \bigcup_{i=1}^{\infty} A_i \subset \bigcup_{i=1}^{\infty} \bigcup_{j=1}^{\infty} B_{i,j}$$

$$\Rightarrow m^*\left(\bigcup_{i=1}^{\infty} A_i\right) \leq \sum_{i,j}^{\infty} |B_{i,j}| \leq \sum_{i=1}^{\infty} m^*(A_i) + 2^{-i} \cdot \varepsilon.$$
$$= \sum_{i=1}^{\infty} m^*(A_i) + \varepsilon.$$

Remark: ①) Countable Subadditivity \Rightarrow finite Subadditivity


In fact, Jordan outer measure satisfies finite Subadditivity but not countable Subadditivity

②) Both Jordan outer measure and Lebesgue outer measure do NOT satisfy finite additivity

Prop. (Translation-invariance)

$$m^*(A + \{x\}) = m^*(A).$$

B_1, B_2, \dots of A

$\Leftrightarrow B_1 + \{x\}, \dots$ of $A + \{x\}$.

Prop: For any elementary set $E \subset R^d$, we have

$$m^*(E) = m(E) \leftarrow \begin{cases} m(E) = \sum_{i=1}^n |B_i|, \text{ where} \\ E = \bigcup_{i=1}^n B_i, \text{ non-overlapping} \end{cases}$$

($m^*(E) \leq m(E)$:)

$$m^*(E) \leq J^*(E) = m(E).$$

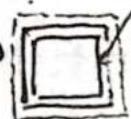
$$(m^*(E) \geq m(E)).$$

prove $m(E) \leq m^*(E) + \varepsilon$. for any ε .

We will use H-B property, (which R satisfies, but Q not)

open cover \leftrightarrow compact set.

for a box



Step 1

Suppose $E = \bigcup_{i=1}^n B_i$, where B_i are non-overlapping.

Shrink B_i to \tilde{B}_i s.t. \tilde{B}_i is compact

Such that $|B_i| \geq |B_i| - \frac{\epsilon}{n}$

$\Rightarrow \tilde{E} = \bigcup_{i=1}^n \tilde{B_i}$ is a compact set, $\tilde{E} \subset E$.

and $m(\tilde{E}) \geq m(E) - \varepsilon$

Step 2

$$\text{st. } \sum_{i=1}^{\infty} |c_i| \leq m^*(E) \uparrow \varepsilon.$$

For any covering of E by boxes $C_1, C_2, C_3, \dots \leftarrow E \subset \bigcup_{i=1}^{\infty} C_i$

We enlarge C_i to a larger open box \tilde{C}_i s.t. $C_i \subset \tilde{C}_i$

and $|c_i| \leq |c_i| + 2^i \cdot \varepsilon$

Then $\bigcup_{i=1}^{\infty} \tilde{C}_i > \bigcup_{i=1}^{\infty} C_i > E > \underline{E} \Rightarrow$ By H-B.

$$\exists \tilde{C}_{k(1)}, \dots, \tilde{C}_{k(N)} \text{ st: } \tilde{E} \subset \bigcup_{i=1}^N \tilde{C}_{k(i)}.$$

$$\begin{aligned} \therefore m(E) &\leq m(\tilde{E}) + \varepsilon \leq m\left(\bigcup_{i=1}^n C_{K(i)}\right) + \varepsilon \leq \sum_{i=1}^n |C_{K(i)}| + \varepsilon \\ &\leq \sum_{i=1}^{\infty} |\tilde{C}_i| + \varepsilon \leq \sum_{i=1}^{\infty} |c_i| + 2\varepsilon \leq m^*(E) + 3\varepsilon. \end{aligned}$$

Def: // we say two boxes B_1 and B_2 are almost disjoint if their interior do not intersect $B_1^\circ \cap B_2^\circ = \emptyset$.

Similarly define 'almost disjoint' for more boxes

Obviously. $m(\bigcup_{i=1}^n B_i) = \sum_{i=1}^n |B_i|$.

if B_i 's are "almost disjoint".

Prop: // Let $A = \bigcup_{i=1}^{\infty} B_i$ be the union of almost disjoint boxes.

Then $m^*(A) = \sum_{i=1}^{\infty} |B_i|$.

Proof. $m^*(A) \leq \sum_{i=1}^{\infty} |B_i|$ by definition (or infinite subadditivity).

Conversely. for any n . $\bigcup_{i=1}^n B_i \subset A$.

So by monotonicity $\sum_{i=1}^n |B_i| = m\left(\bigcup_{i=1}^n B_i\right) = \underline{m^*\left(\bigcup_{i=1}^n B_i\right)} \leq m^*(A)$.

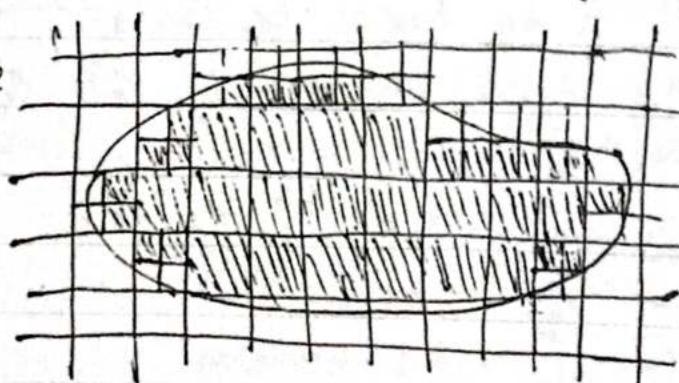
(Let $n \rightarrow \infty$.)

or disjoint semi-open and semi-closed
boxes

Thm. (Structure of open sets in R^d).

Any open set $A \subset R^d$ can be expressed as the union of countable many almost disjoint (closed) boxes.

Proof



$$\Leftrightarrow m^*(A) = \sum_{i=1}^{\infty} |B_i| \text{ for any open set } A$$

Consider dyadic closed boxes of the form

$$Q = \left[\frac{i_1}{2^n}, \frac{i_1+1}{2^n}\right] \times \left[\frac{i_2}{2^n}, \frac{i_2+1}{2^n}\right] \times \cdots \times \left[\frac{i_d}{2^n}, \frac{i_d+1}{2^n}\right]$$

$i_j \in \mathbb{Z}$, $n \in \mathbb{N}$.

let $A = \{\mathcal{Q} \mid \mathcal{Q} \text{ as above. } \mathcal{Q} \subset A\}$.

Claim // $\bigcup_{\mathcal{Q} \in A} \mathcal{Q} = A$.

- $\bigcup_{\mathcal{Q} \in A} \mathcal{Q} \subset A$ obvious.
- For any $x \in A$. A open thus, exists a small ball $B_r(x) \subset A$. take n large s.t. $\frac{1}{2^n} < r_0$. Then we can find a $\mathcal{Q} \ni x$ s.t. $\mathcal{Q} \subset B_r(x) \subset A$. $\therefore A \subset \bigcup_{\mathcal{Q} \in A} \mathcal{Q}$.

$$\therefore \bigcup_{\mathcal{Q} \in A} \mathcal{Q} = A.$$

Observation: For any such closed boxes $\mathcal{Q}_1, \mathcal{Q}_2$.
either they are almost disjoint

or one is contained in the other

(let $A^* = \{\mathcal{Q} \mid \mathcal{Q} \subset A, \mathcal{Q} \text{ is not contained in a large } \mathcal{Q}' \text{, not } \subset A\}$)

Then $A = \bigcup_{\mathcal{Q} \in A^*} \mathcal{Q}$. \leftarrow countable union of almost disjoint closed boxes.

Prop. (outer regularity) For any $A \subset \mathbb{R}^d$ $m^*(A) = \inf_{\substack{U \supset A \\ U \text{ open}}} m^*(U)$.

Proof. By monotonicity. $A \subset U \Rightarrow m^*(A) \leq m^*(U)$
 $\therefore m^*(A) \leq \inf_{U \supset A} m^*(U)$.

To prove the converse. W.L.O.G. assume $m^*(A) < \infty$.
 $\left(\inf_{U \supset A \text{ open}} m^*(U) \leq m^*(A) \right)$

By definition of Lebesgue outer measure, \checkmark one can find boxes B_1, B_2, \dots

s.t. $A \subset \bigcup_{i=1}^{\infty} B_i$ and $\sum_{i=1}^{\infty} |B_i| < m^*(A) + \varepsilon$.

Enlarge each B_i to \tilde{B}_i . \tilde{B}_i is open box
and $|\tilde{B}_i| \leq |B_i| + 2^{-i}\varepsilon$. Let $U = \bigcup_{i=1}^{\infty} \tilde{B}_i$

thus, $A \subset \bigcup_{i=1}^{\infty} B_i \subset \bigcup_{i=1}^{\infty} \tilde{B}_i = U$

$$\Rightarrow \inf_{\substack{V \supset A \\ V \text{ open}}} m^*(U) \leq m^*(U_1) \leq \sum_{i=1}^{\infty} |\tilde{B}_i| \leq \sum_{i=1}^{\infty} |B_i| + 2^{-i}\varepsilon \leq m^*(A) + 2\varepsilon.$$

Let $\varepsilon \rightarrow 0$. we get $\inf_{\substack{V \supset A \\ V \text{ open}}} m^*(U) \leq m^*(A)$

Lee 4

Additivity: (In condition of not overlapping)

\rightarrow If $d(A_1, A_2) > 0$, then $m^*(A_1 \cup A_2) = m^*(A_1) + m^*(A_2)$
(Stranger than $A_1 \cap A_2 = \emptyset$)

$\rightarrow m^*(\bigcup_{i=1}^{\infty} B_i) = \sum_{i=1}^{\infty} |B_i|$ for almost disjoint boxes

\rightarrow Elementary set In Jordan measurable, $J(A)$ exists $\Leftrightarrow \forall \varepsilon > 0 \exists F \in \mathcal{F}$ s.t. $J^*(F \setminus A) < \varepsilon$.

Def: || A set $A \subset R^d$ is Lebesgue measurable if $\forall \varepsilon > 0 \exists$ open set $V \supset A$ s.t. $m^*(V \setminus A) < \varepsilon$.

- If A is Lebesgue measurable, we write $m(A) = m^*(A)$. and call it Lebesgue measure of A .
- We denote the set of all Lebesgue measurable sets by \mathcal{L} .

"Lebesgue measurable" \approx nearly open

If we want to compare A_1, A_2 .

we can consider $A_1 \Delta A_2 = (A_1 \setminus A_2) \cup (A_2 \setminus A_1)$.

As set today

↗ difference between Lebesgue outer measure and Lebesgue measurable

$\nexists \varepsilon > 0 \exists V \supset A$ open s.t. $m^*(V) - m^*(A) < \varepsilon$.

$\left\{ \begin{array}{l} \nexists \varepsilon > 0 \exists V \supset A \text{ open s.t. } m^*(V \setminus A) < \varepsilon. \end{array} \right.$

Example: Any countable set is Lebesgue measurable
with Lebesgue measurable = 0.

- If $m^*(A) = 0$, then A is Lebesgue measurable, with $m(A) = 0$.
 $\forall \epsilon > 0 \exists V \supset A$ s.t. $m^*(V) < \epsilon$ $m^*(V \setminus A) < m^*(V)$

Example: Any open set is Lebesgue measurable.
Any box is Lebesgue measurable.

Prop. If $A_i \in \mathcal{L}$, then $\bigcup_{i=1}^{\infty} A_i \in \mathcal{L}$.

Proof: $[\epsilon = \frac{\epsilon}{2} + \frac{\epsilon}{4} + \dots + \frac{\epsilon}{2^n} + \dots]$.

For any i , find an open set $U_i \supset A_i$ s.t. $m^*(U_i \setminus A_i) < \frac{\epsilon}{2^i}$.

Let $U = \bigcup_{i=1}^{\infty} U_i$. Then U is open $U \supset \bigcup_{i=1}^{\infty} A_i$

$$m^*(U \setminus A) \leq m^*\left(\bigcup_{i=1}^{\infty} (U_i \setminus A_i)\right) \leq \sum_{i=1}^{\infty} m^*(U_i \setminus A_i) < \epsilon.$$

* $\boxed{U \setminus A \subset \bigcup_{i=1}^{\infty} (U_i \setminus A_i)}$
 $A = \bigcup_{i=1}^{\infty} A_i$

• Any elementary is Lebesgue measurable.

Prop: Any compact set is Lebesgue measurable.

Proof: Let $A \subset \mathbb{R}^d$ be compact

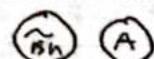
By def, we can find open set U . s.t. $U \supset A$.

$$\text{and } m^*(U) \leq m^*(A) + \epsilon \quad \text{+ additivity} = m^*(A) = m(A).$$

Since $U \setminus A$ is open, we can write $U \setminus A = \bigcup_{i=1}^{\infty} B_i$

(B_i are almost disjoint closed boxes).

Set $\widehat{B}_n = \bigcup_{i=1}^n B_i$. Then \widehat{B}_n is closed, and $\widehat{B}_n \cap A = \emptyset$.
 $\Rightarrow d(\widehat{B}_n, A) > 0$.



$$\text{So, } m^*(A) + m^*(\widehat{B}_n) = m^*(A \cup \widehat{B}_n) \leq m^*(A \cup (U \setminus A)) = m^*(U) \leq m^*(A)$$

Note.. A compact $\Rightarrow A$ is contained in a large box

$$\Rightarrow m^*(A) < \infty. \text{ So } m^*(\tilde{B}_n) < \varepsilon.$$

$$\Rightarrow m^*(\tilde{B}_n) = \sum_{i=1}^n m^*(B_i) = \sum_{i=1}^n |B_i| < \varepsilon \xrightarrow{n \rightarrow \infty} \sum_{i=1}^{\infty} |B_i| < \varepsilon, \text{ i.e. } \\ (\text{for all } n).$$

$$m^*(U \setminus A) < \varepsilon.$$

Cor. Any closed set is Lebesgue measurable.

Proof: A closed $\Rightarrow A_n := A \cap \overline{B(0,n)}$ is compact (for all n)

and $A = \bigcup_{n=1}^{\infty} A_n$. (develop compact sets to closed sets)

Prop 1.

Prop: If $A \in \mathcal{L}$, then $A^c = \mathbb{R}^d \setminus A \in \mathcal{L}$.

Proof: Take open sets $U_n \supset A$, s.t: $m^*(U_n \setminus A) < \frac{1}{n}$.

Let $F_n = U_n^c$, and $F = \bigcup_{n=1}^{\infty} F_n$ Then $F \in \mathcal{L}$.

Since $A^c \setminus F \subset A^c \setminus F_n$, and $m^*(A^c \setminus F_n) = m^*(U_n \setminus A) < \frac{1}{n}$.

$$\Rightarrow m^*(A^c \setminus F) \leq m^*(A^c \setminus F_n) < \frac{1}{n}. \xrightarrow{n \rightarrow \infty} m^*(A^c \setminus F) = 0 \Rightarrow A^c \setminus F \in \mathcal{L}$$

$$\Rightarrow \begin{cases} F \in \mathcal{L} \\ A^c \setminus F \in \mathcal{L} \end{cases} \Rightarrow A^c \in \mathcal{L}.$$

$$\begin{aligned} A_1 \setminus A_2 &= A_1 \cap A_2^c \\ &= A_2^c \cap (A_1^c)^c = A_2^c \setminus A_1^c. \end{aligned}$$

Cor II $A_1 \in \mathcal{L} \Rightarrow \bigcap_{i=1}^{\infty} A_i \in \mathcal{L}$.

Proof $\left(\bigcap_{i=1}^{\infty} A_i\right)^c = \left(\bigcup_{i=1}^{\infty} A_i^c\right) \in \mathcal{L}$.

$$\bigcap_{i=1}^{\infty} A_i^c \in \mathcal{L}$$

- For Sets (countable many closed sets' union) $\in \mathcal{L}$

- 6 sets $\in \mathcal{L}$. $G \& F = \bigcap_{i=1}^{\infty} A_i$. A_i 's are open sets.

Cor II $A_1, A_2 \in \mathcal{L} \Rightarrow A_1 \setminus A_2 \in \mathcal{L}$.

Proof: $A_1 \setminus A_2 = A_1 \cap A_2^c$.

Prop: $A \in \mathcal{L} \Leftrightarrow \forall \varepsilon > 0. \exists$ closed set $F \subset A$ s.t. $m^*(A \setminus F) < \varepsilon$.

Proof: $A \in \mathcal{L} \Leftrightarrow A^c \in \mathcal{L} \Leftrightarrow \exists U \supset A^c$ s.t. $m^*(U \setminus A^c) < \varepsilon$.

$$\Leftrightarrow m^*(A \setminus U^c) < \varepsilon$$

* closed and $A \supset U^c$.

→ finite additivity.

Theorem (Countable additivity of Lebesgue measure)

$$A_i \in \mathcal{L}, \text{ disjoint} \Rightarrow m\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} m(A_i).$$

Proof: • Easy condition:

Assume all A_i 's are compact. Then (finite additivity for compact sets)

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

$$\text{Let } n \rightarrow \infty \text{ we get } m\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} m(A_i).$$

• Middle case. Assume all A_i 's are bounded.

$$A_i \in \mathcal{L} \Rightarrow \exists \text{ closed } F_i \subset A_i. \text{ s.t. } m(A_i \setminus F_i) < \varepsilon \cdot 2^{-i}$$

(compact)

Note: A_i 's are disjoint. $\Rightarrow F_i$'s are disjoint.

$$\sum_{i=1}^{\infty} m(A_i) \leq \sum_{i=1}^{\infty} (m(F_i) + m(A_i \setminus F_i)) \leq \sum_{i=1}^{\infty} m(F_i) + \varepsilon.$$

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

$$\text{Let } \varepsilon \rightarrow 0 \Rightarrow m\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} m(A_i)$$

General Case Only assume $A_i \in \mathcal{L}$.

Let $C_n = B(0, n) \setminus B(0, n-1)$.

Let $D_{i,n} = A_i \cap C_n$ They are bounded.

$$\text{Then } \sum_{i=1}^{\infty} m(A_i) = \left(\sum_{i=1}^{\infty} \sum_{n=1}^{\infty} m(D_{i,n})\right) = m\left(\bigcup_{i,n} D_{i,n}\right) = m\left(\bigcup_{i=1}^{\infty} A_i\right).$$

Countable

Cor [Monotone convergence of measure]

Suppose $A_1 \subset A_2 \subset \dots \subset A_i \in \mathcal{L}$ Then

$$m\left(\bigcup_{i=1}^{\infty} A_i\right) = m\left(\lim_{n \rightarrow \infty} A_n\right) = \lim_{n \rightarrow \infty} m(A_n).$$

Proof: Let $A_1' = A_1, A_2' = A_2 - A_1, \dots, A_k' = A_k - A_{k-1}, \dots$

$\Rightarrow A_i'$ are disjoint and $\in \mathcal{L}$.

$$\Rightarrow m\left(\bigcup_{i=1}^{\infty} A_i\right) = m\left(\bigcup_{i=1}^{\infty} A_i'\right) = \sum_{i=1}^{\infty} m(A_i') = \sum_{i=1}^{\infty} (m(A_i) - m(A_{i-1})).$$

$A_i = A_i' \cup A_{i-1}$

$$= \lim_{n \rightarrow \infty} m(A_n).$$

Lee 5

Last time

- A is Lebesgue measurable $\Leftrightarrow \forall \varepsilon > 0 \exists$ open set $U \supset A$ st. $m^*(U \setminus A) < \varepsilon$
- $\Leftrightarrow \forall \varepsilon > 0 \exists$ closed set $F \subset A$ st. $m^*(A \setminus F) < \varepsilon$.

① $\emptyset \in \mathcal{L}$

② $A \in \mathcal{L} \Rightarrow A^c \in \mathcal{L}$

③ $A \in \mathcal{L} \Rightarrow \bigcup_{n=1}^{\infty} A_n \in \mathcal{L}$

$\bigcap_n A_n \in \mathcal{L}$

\mathcal{L} is a σ -Algebra

$$B = \bigcap_{\Sigma} \sum_{\substack{\text{OCRd. or open} \\ \Sigma \text{ is a } \sigma\text{-algebra}}}$$

$\forall B \in B$.

B is a Borel set.

Examples • countable sets

- $m^*(A) = 0$ (we call it null sets).
- open, closed, compact $\in \mathcal{L}$.
- Fo. Gg. $\in \mathcal{L}$.

Countable
 \bigcup for closed
sets

Countable
 \bigcap for open
sets

$\mathbb{Q} \in \mathcal{L}$

$\mathcal{T} \subset \mathcal{L}$

Additivity (most significant property compared with m^* .)

Suppose $A_n \in \mathcal{L}$ are disjoint

$$\text{then } m\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} m(A_n). \quad \text{If } m(A_n) \text{ can be } +\infty.$$

Prop || T.F.A.E

(1) $A \in \mathcal{L}$.

(2) $A = G \setminus N$, where G is G_δ set, N is a null set

(3) $A = F \cup N$, where F is a F_σ set, N is a null set.

Weaker than E . F open. Stronger than ϵ . F closed.

Proof: (1) \Rightarrow (2) For any n . Let U_n be an open set

$$\text{s.t. } U_n \supset A, m^*(U_n \setminus A) < \frac{1}{n}.$$

Let $G = \bigcap_{n=1}^{\infty} (U_n \cap G_\delta)$ and $G \supset A$.

Note. Let $N = G \setminus A$. Then $N \subset U_n \setminus A$ (for each n)

By monotonicity of m^* , $m^*(N) \leq m^*(U_n \setminus A) < \frac{1}{n}$.

(as $n \rightarrow \infty$, $m^*(N) = 0$. i.e. N is a null set)

(2) \Rightarrow (1) $G \in \mathcal{L}, N \in \mathcal{L}, G \setminus N \in \mathcal{L}$.

(1) \Rightarrow (3) For any n let F_n be an closed set.

$$\text{s.t. } F_n \subset A, m^*(A \setminus F_n) < \frac{1}{n}$$

Let $F = \bigcup_{n=1}^{\infty} F_n \in F_\sigma$. and $F \subset A$.

Note. Let $N = A \setminus F$. Then $N \subset A \setminus F_n$

$$m^*(A \setminus F) \leq m^*(A \setminus F_n) < \frac{1}{n}. \text{ (as } n \rightarrow \infty \text{, } m^*(A \setminus F) = 0\text{.)}$$

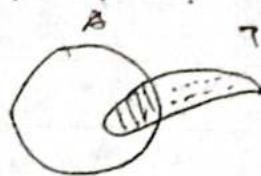
i.e. $N = A \setminus F$ is a null set.

(3) \Rightarrow (1) $A = F \cup N, F \in \mathcal{L}, N \in \mathcal{L} \therefore A \in \mathcal{L}$.

Other way to prove (1) \Leftrightarrow (3).

$$A = G \setminus N = G \cap N^c \quad A^c = (G^c \cup N) \cap A^c \subset F_\sigma \text{ set}$$

Prop: Suppose $A \subset \mathbb{R}^d$ is Lebesgue measurable
Then for any $T \subset \mathbb{R}^d$.
 $m^*(T) = m^*(T \cap A) + m^*(T \setminus A)$



Proof. First assume T is open, then by additivity of m .

$$m(T) = m(T \cap A) + m(T \setminus A). \quad \textcircled{1}$$

Now for general $T \subset \mathbb{R}^d$, we take an open set $U \supset T$.

st: $m^*(U) \leq m^*(T) + \varepsilon$ definition of m^* . Cannot write $m^*(U \setminus T) < \varepsilon$ here

$$\text{Then } m^*(T) \leq m^*(T \cap A) + m^*(T \setminus A)$$

$$\leq m^*(U \cap A) + m^*(U \setminus A) \leftarrow \text{using } \textcircled{1}$$

$$= m^*(U) \leq m^*(T) + \varepsilon \quad (\text{let } \varepsilon \rightarrow 0)$$

$$\Rightarrow m^*(T) = m^*(T \cap A) + m^*(T \setminus A).$$

Thm. T.F.A.E. (Carathéodory Criteria).

$$(1) A \in \mathcal{L}$$

$$(2) \text{ For any box } B. \quad |B| = m^*(B \cap A) + m^*(B \setminus A)$$

$$(3) \text{ For any } E \in \mathcal{E} \quad m(E) = m^*(E \cap A) + m^*(E \setminus A)$$

$$(4) \text{ For any set } T. \quad m^*(T) = m^*(T \cap A) + m^*(T \setminus A)$$

definition of measurable in textbook

This is used to define measurable sets in abstract outer measure space (without open sets).

$$(1) \Rightarrow (4) \vee (4) \Rightarrow (1)$$

$$(2) \Rightarrow (1) \quad (2) \Rightarrow (3), \quad (2)+(3) \Rightarrow (1)$$

$$(2) \Rightarrow (3) \text{ Suppose } E \in \mathcal{E} \text{ Then } E = \bigcup_{i=1}^n B_i$$

$$\text{So } m(E) = \sum_{i=1}^n |B_i| = \sum_{i=1}^n m^*(B_i \cap A) + \sum_{i=1}^n m^*(B_i \setminus A)$$

$$m(E) \leq m^*(E \cap A) + m^*(E \setminus A) \leq \sum m^*(B_i \cap A) + \sum m^*(B_i \setminus A)$$

First assume $m^*(A) < \infty$

$$(2) \\ (3) \Rightarrow (4)$$

Take disjoint boxes B_i 's s.t. $A \subset \bigcup_{i=1}^{\infty} B_i$
 $\sum_{i=1}^{\infty} |B_i| \leq m^*(A) + \varepsilon.$

Enlarge each B_i to an open $\tilde{B}_i \supset B_i$.

$$\text{s.t. } |\tilde{B}_i| < |B_i| + \frac{\varepsilon}{2\pi}$$

Then $\bigcup_{i=1}^{\infty} \tilde{B}_i = U$ is a open set, and $U \supset A$. and.

$$\begin{aligned} m^*(A) &= m^*(A \cap U) \\ &= m^*(U \setminus (U \cap A)) \\ &= m^*(\bigcup_{i=1}^{\infty} (\tilde{B}_i \setminus A)) + m^*(U \setminus \bigcup_{i=1}^{\infty} \tilde{B}_i) \\ &\Rightarrow m^*(U \setminus A) \leq 2\varepsilon. \end{aligned}$$

Second, assume $m^*(A) = \infty$

Let $D_n = b_n \times nR^d$, centered at 0. Side length is n .

Then $A \cap D_n = A_n \rightsquigarrow m^*(A) \leq m^*(D_n) < \infty$.

$$\text{and } A = A \cap R^d = A \cap \left(\bigcup_{i=1}^{\infty} D_i \right) = \bigcup_{i=1}^{\infty} (A \cap D_i) = \bigcup_{i=1}^{\infty} A_n$$

If we can prove each $A_n \in \mathcal{L}$ the $A = \bigcup_{i=1}^{\infty} A_n \in \mathcal{L}$

By 1st part, we only need to prove

$$|B| = m^*(B \cap A_n) + m^*(B \setminus A_n) \quad (\text{for any } B)$$

$$\text{We need } B \setminus A_n = B \cap A_n^c = B \cap (A \cap D_n)^c = B \cap (A^c \cup D_n^c).$$

$$= (B \cap A^c) \cup (B \cap D_n^c) = (B \cap A^c \cap D_n) \cup (B \cap A^c \cap D_n^c) \cup (B \cap D_n^c)$$

$$\begin{aligned} \text{Note } |B| &\leq m^*(B \cap A_n) + m^*(B \setminus A_n) \\ &\leq m^*(B \cap D_n \cap A) + m^*(B \cap A^c \cap D_n) + m^*(B \cap D_n^c). \end{aligned}$$

↑? $m^*(B \cap D_n \cap A)$ $m^*(B \cap D_n)$ $m^*(B \cap D_n^c)$

$$\begin{aligned} &= m^*((B \cap D_n) \cap A) + m^*((B \cap D_n) \setminus A) + m^*(B \cap D_n^c) \\ &= m^*(B \cap D_n) + m^*(B \setminus D_n) = m(B) = |B|. \end{aligned}$$

$$\Rightarrow |B| = m^*(B \cap A_n) + m^*(B \setminus A_n).$$

Prop. Suppose $A \subset \mathbb{R}^d$ is Lebesgue measurable $m(A) > 0$. ($m^*(A) > 0$)
 Then $\forall 0 < t < 1 \exists$ box B s.t. $m(B \cap A) > t|B|$. ($m^*(B \cap A) > t|B|$)
 (Some convergence property for measurable sets).

Proof. First assume $m(A) < +\infty$.

$\frac{(1-t)m(A)}{t} > 0$. One can take open set $U \supset A$. s.t. $m^*(U \setminus A) < \epsilon$.

By structure theorem, we can write $U = \bigcup_{n=1}^{\infty} B_n$. almost disjoint boxes.

WANT $\exists k$ s.t. $t|B_k| < m(B_k \cap A)$.

By contradiction. Assume $m(B_n \cap A) \leq t|B_n| \quad (\forall n)$

$$m(A) \leq \sum_{n=1}^{\infty} m(A \cap B_n) \leq t \sum_{n=1}^{\infty} |B_n| = t m(U) \leq t(m(U \setminus A) + m(A))$$

$$\leq t m(A) + t\epsilon < m(A)$$

↓ to get contradiction.

$$\text{need } \epsilon < \frac{(1-t)m(A)}{t}$$

For $m(A) = +\infty$.

Then $\exists n$ s.t. $A_n = A \cap B(0, n)$ s.t. $m(A_n) > 0$ (monotone convergence)

Apply 1st part to A_n .

Prop. (Steinhaus) || Suppose $A \subset \mathbb{R}$ $m(A) > 0$. Then

$$A - A := \{x - y \mid x, y \in A\}.$$

|| Contains a small box B centered at 0.

Thm (Uniqueness of Lebesgue measure)

Suppose $\tilde{m}: \mathcal{L} \rightarrow [0, +\infty]$ satisfies

(1) $\tilde{m}(\emptyset) = 0$

(2) For disjoint $A, B \in \mathcal{L}$ one has $\tilde{m}(\bigcup_{n=1}^{\infty} A_n) = \sum_{n=1}^{\infty} \tilde{m}(A_n)$.

(3) $\tilde{m}(x + A) = \tilde{m}(A)$

(4) $\tilde{m}([0, 1]^d) = 1$.

Then $\tilde{m} = m$ i.e. $\tilde{m}(A) = m(A)$

Proof By lecture 2 (P Set 1-2, Problem 2) we see that for $E \in \mathcal{E}$, then $\tilde{m}(E) = m(E)$

- By additivity and non-negativity, we have } monotonicity
stronger than }
Sub-additivity:

Step 1: $m(A) = 0 \Rightarrow \tilde{m}(A) = 0$.

Proof: Cover A by boxes B_n 's s.t. $A \subset \bigcup_{n=1}^{\infty} B_n$ and $\sum_{n=1}^{\infty} |B_n| < \varepsilon$.

$$\text{Then } \tilde{m}(A) \leq \tilde{m}\left(\bigcup_{n=1}^{\infty} B_n\right) \leq \sum_{n=1}^{\infty} \tilde{m}(B_n) = \sum_{n=1}^{\infty} m(B_n) < \varepsilon.$$

Let $\varepsilon \rightarrow 0$. we get $\tilde{m}(A) = 0$.

Step 2: If U is open $m(U) < \infty$, then $\tilde{m}(U) = m(U)$.

By structure theorem. $U = \bigcup_{n=1}^{\infty} B_n$. B_n 's are almost disjoint closed boxes
 $\Rightarrow B_n$ are disjoint open boxes.

Let $F = U \setminus \bigcup_{n=1}^{\infty} B_n \in \mathcal{L}$

$$\text{Then } m(F) = m(U) - m\left(\bigcup_{n=1}^{\infty} B_n\right) = \sum_{n=1}^{\infty} |B_n| - \sum_{n=1}^{\infty} |B_n| = 0.$$

$$\Rightarrow \tilde{m}(F) = 0.$$

$$\begin{aligned} \text{So } \tilde{m}(U) &= \tilde{m}(F) + \tilde{m}\left(\bigcup_{n=1}^{\infty} B_n\right) = \tilde{m}\left(\bigcup_{n=1}^{\infty} B_n\right) = \sum_{n=1}^{\infty} \tilde{m}(B_n) \\ &= \sum_{n=1}^{\infty} m(B_n) = m(U) - m(F) = m(U) \end{aligned}$$

Step 3: $\tilde{m}(A) \leq m(A)$ (for $A \in \mathcal{L}$).

• It is trivially true for $m(A) = +\infty$

• Now assume $m(A) < \infty$.

Choose open set $V \supset A$ s.t. $m(V \setminus A) = m(V) - m(A) < \varepsilon$.

So. $\tilde{m}(A) \leq \tilde{m}(U) = m(U) < m(A) + \varepsilon$. $\varepsilon \rightarrow 0$

$$\Rightarrow \tilde{m}(A) \leq m(A).$$

(thought of A^c)

Step 4: If $A \in \mathcal{L}$ is bounded, then $\tilde{m}(A) = m(A)$.

Take a box B s.t. $A \subset B$.

By additivity. $\tilde{m}(B \setminus A) = |B| - m(A)$.

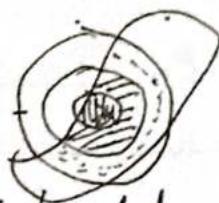
$$\text{Then } \tilde{m}(A) \leq m(A) = m(B) - m(B \setminus A) \leq \tilde{m}(B) - \tilde{m}(B \setminus A) = \tilde{m}(A)$$

Step 5. For general $A \in \mathcal{L}$.

Let $C_n = B(0, n) \setminus B(0, n-1)$ $A_n = A \cap C_n$

Then $A = \bigcup_{n=1}^{\infty} A_n$. disjoint union. each A_n is bounded.

So $\tilde{m}(A) = \sum_{n=1}^{\infty} \tilde{m}(A_n) = \sum_{n=1}^{\infty} m(A_n) = m(A)$.



Today Lee 6.

Recall (Lee 1).

Lebesgue's idea of integration

$$\int_a^b f(x) dx = \int_0^\infty m(\{x | f(x) > t\}) dt$$

$f(x) \geq 0$

Def: We say a variable function f is (Lebesgue) measurable if for any $t \in \mathbb{R}$, the set $\{x | f(x) > t\}$ is Lebesgue measurable.

Remarks:

i) We allow the domain of f to be a set A , instead of \mathbb{R}^n .

In this case, we require $A \in \mathcal{L}$.

Example: For any set $A \subset \mathbb{R}^d$ consider the characteristic function

$$\chi_A(x) = \begin{cases} 1 & x \in A \\ 0 & x \notin A \end{cases}$$

Then χ_A is measurable $\Leftrightarrow A \in \mathcal{L}$.

2) we allow the value of f to be $\pm\infty$.

In which case we say f is \mathbb{R}^* -valued.
 $\mathbb{R}V\{\pm\infty\}$

WARNING: $(+\infty) + (-\infty)$ makes no sense.

$$f(x) = x^2$$

In the case all values of f are in \mathbb{R} . we say f is finite-valued or \mathbb{R} -valued.

$\begin{cases} \text{finite-valued} \\ \text{bounded function} \end{cases}$

3) We say two functions f and g ^{are} equal almost everywhere.
 (or a.e. equal).

If the set $\{x : f(x) \neq g(x)\}$ is a null set

Example: $X_\alpha = 0$. a.e.

More generally we say a proposition P holds a.e.

If those points where P fails form a null set

Prop: Let $f: A \subset \mathbb{R}^d \rightarrow \mathbb{R}$. where $A \in \mathcal{L}$.

T.T.A.E.

\Leftrightarrow definition of measurable function

(1) $\forall t \in \mathbb{R}$ the set $\{x | f(x) > t\} \in \mathcal{L}$.

(2) $\dots \dots \dots \geq \dots \dots$

(3) $\dots \dots \dots \leq \dots \dots$

(4) $\dots \dots \dots \leq \dots \dots$

Proof: (1) \Leftrightarrow (4). $\{x | f(x) > t\}^c = \{x | f(x) \leq t\}$

(2) \Leftrightarrow (3) $(C, >)$

(2) \Rightarrow (1). $\{x | f(x) > t\} = \bigcup_{n=1}^{\infty} \{x | f(x) \geq t + \frac{1}{n}\}$. \star some set is
 a way to prove measurable.

(4) \Rightarrow (3) SAME method

$$\begin{array}{ccc} (1) & \Leftrightarrow & (4) \\ \uparrow & & \downarrow \\ (2) & \Leftrightarrow & (3) \end{array}$$

Remark: We are not saying for a given t_0
 $\{x \mid f(x) > t_0\} \in \mathcal{L} \Leftrightarrow \{x \mid f(x) \geq t_0\} \in \mathcal{L}$.

② (P Set Today)

(5) If open set $U \subset \mathbb{R}$, $f^{-1}(U) \in \mathcal{L}$.

(6) If closed set $F \subset \mathbb{R}$, $f^{-1}(F) \in \mathcal{L}$.

Cor: If f is Lebesgue measurable, then

for any t , the level set

$$\begin{aligned} f^{-1}(t) &= \{x \mid f(x) = t\}, \text{ is Lebesgue measurable.} \\ &= \{x \mid f(x) \geq t\} \cap \{x \mid f(x) \leq t\}. \end{aligned}$$

Examples of measurable functions.

(1) χ_A , where $A \in \mathcal{L}$.

(2) Any monotone function is measurable. (\mathbb{R})

(3) Any continuous function is measurable

\Leftrightarrow for any open set U , $f^{-1}(U)$ is open. $\in \mathcal{L}$.

Remark ② (5).

(4) ϕ of where $\phi: \mathbb{R} \rightarrow \mathbb{R}$ is continuous* and f is Lebesgue measurable.

$$(\phi \circ f)^{-1}(U) = \underbrace{f^{-1}}_{\text{open}} \circ \underbrace{\phi^{-1}(U)}_{\text{open}} \in \mathcal{L}.$$

WARNING: The composition of two measurable functions need not to be measurable function.

(5) Prop: If f, g are \mathbb{R} -valued measurable function.

then " $f + g$, $f \cdot g$ are measurable
 Algebraic Structure (except for \mathbb{R}^* -valued functions here).

Proof 1) If $c > 0$, $\{x | cf(x) > t\} = \{x | f(x) > \frac{t}{c}\}$.

$$c < 0$$

$$c = 0 \quad cf = 0.$$

(2) [Trick: decomposition a set to a countable union]

Suppose $f(x) + g(x) > t$.

$$\Leftrightarrow f(x) > t - g(x)$$

$$\Leftrightarrow \exists r \in \mathbb{Q} \text{ s.t. } f(x) > r > t - g(x).$$

$$\Leftrightarrow \exists r \in \mathbb{Q} \text{ s.t. } f(x) > r \text{ and } g(x) > t - r.$$

So:

$$\{x | f(x) + g(x) > t\} = \bigcup_{r \in \mathbb{Q}} \left(\{x | f(x) > r\} \cap \{x | g(x) > t - r\} \right).$$

(3) First note: f is measurable $\Rightarrow f^2$ is measurable.

$$\{x | f^2(x) > t\}_{t > 0} = \{x | |f(x)| > \sqrt{t}\} = \{x | f(x) > \sqrt{t}\} \cup \{x | f(x) < -\sqrt{t}\}$$

Now suppose f, g measurable. Then

$$fg = \frac{1}{2}((f+g)^2 - f^2 - g^2) \text{ is measurable.}$$

(6)

Thm || If f_n are measurable functions.

and $\lim_{n \rightarrow \infty} f_n = f$ then f is measurable.

Remark: we know that limit of continuous functions need not be continuous.

So measurable functions are much "nicer" than continuous functions under taking limit

Proof: Suppose

$$f(x) = \lim_{n \rightarrow \infty} f_n(x) > t. \Rightarrow \exists k \text{ s.t. } \lim_{n \rightarrow \infty} f_n(x) > t + k.$$

$\Rightarrow \exists k, \exists N \text{ s.t. } \forall n > N f_n(x) > t + \frac{1}{k}$.

$$\text{So } \{x \mid f(x) > t\} = \bigcup_k \{x \mid \lim_{n \rightarrow \infty} f_n(x) > t + \frac{1}{k}\}.$$

$$= \bigcup_{k=1}^{\infty} \bigcap_{N=1}^{\infty} \bigcup_{n>N} \{x \mid f_n(x) > t + \frac{1}{k}\} \in \mathcal{L} \quad \left. \begin{array}{l} \exists \rightsquigarrow V \\ \forall \rightsquigarrow U \end{array} \right.$$

In fact, we don't need to assume f_n to be convergent.

Prop: Let f_n be any sequence of measurable functions
 then $\begin{cases} \limsup_{n \rightarrow \infty} f_n & \liminf_{n \rightarrow \infty} f_n \\ \sup_n f_n & \inf_n f_n \end{cases}$

are all measurable functions

for a fixed x , $\sup_n f_n(x) > t \Leftrightarrow \exists n \text{ s.t. } f_n(x) > t$.

$$\{x \mid \sup_n f_n(x) > t\} = \bigcup_n \{x \mid f_n(x) > t\}.$$

$$\limsup_{n \rightarrow \infty} f_n(x) > t \Rightarrow \exists k. \limsup_{n \rightarrow \infty} f_n(x) > t + \frac{1}{k}.$$

$$\Rightarrow \exists k. \forall N \exists n > N. \text{ s.t. } f_n(x) > t + \frac{1}{k}.$$

$$\{x \mid \limsup_{n \rightarrow \infty} f_n(x) > t\} = \bigcup_{k=1}^{\infty} \bigcap_{N=1}^{\infty} \bigcup_{n=N}^{\infty} \{x \mid f_n(x) > t + \frac{1}{k}\}.$$

Cor. f is measurable $\Rightarrow |f|$ is measurable.

$$|f| = \sup \{f_+ - f_-\}$$

Any measurable function is the limit of a sequence
of simple function

Lec 7

Today: Convergent sequence of measurable functions

1 Convergence in measure

Let's take a closer look at a.e. convergence.

$$\begin{aligned} (\lim_{n \rightarrow \infty} f_n(x) \neq f(x)) &\Leftrightarrow \exists \varepsilon > 0 \text{ s.t. } \forall N \in \mathbb{N}, \exists k > N \text{ s.t. } |f_k(x) - f(x)| > \varepsilon. \\ (\lim_{n \rightarrow \infty} f_n(x) = f(x)) &\Leftrightarrow \forall \varepsilon > 0 \exists N \in \mathbb{N} \text{ s.t. } \forall k > N, |f_k(x) - f(x)| < \varepsilon. \end{aligned}$$

Let $A_k = \{x \in A : |f_k(x) - f(x)| > \varepsilon\}$.

$$\left(\lim_{n \rightarrow \infty} f_n(x) \neq f(x) \right) \Leftrightarrow \exists \varepsilon > 0, x \in \bigcap_{N=1}^{\infty} \bigcup_{k=N}^{\infty} A_k(\varepsilon).$$

$$\begin{aligned} f_n \rightarrow f \text{ a.e.} &\Leftrightarrow m(\{x : \lim_{n \rightarrow \infty} f_n(x) \neq f(x)\}) = 0 \Leftrightarrow m(\{x : \exists \varepsilon > 0 \text{ s.t. } x \in \bigcup_{k=N}^{\infty} A_k(\varepsilon)\}) = 0 \\ &\Leftrightarrow m(\bigcup_{\varepsilon} \bigcup_{N=1}^{\infty} \bigcup_{k=N}^{\infty} A_k(\varepsilon)) = 0. \Leftrightarrow \boxed{\forall \varepsilon > 0, m(\limsup_{n \rightarrow \infty} A_n(\varepsilon)) = 0} \end{aligned}$$

infinite union

Def: For any sequence A_k of sets we define its upper limit

$$\limsup_{k \rightarrow \infty} A_k = \bigcap_{N=1}^{\infty} \bigcup_{k=N}^{\infty} A_k$$

Prop: $\limsup_{n \rightarrow \infty} A_k = \{x : x \in A_k \text{ for infinitely many } A_k\}'$

$$\begin{aligned} \text{proof. } X = \bigcap_{N=1}^{\infty} \bigcup_{k=N}^{\infty} A_k &\Leftrightarrow \forall N, x \in \bigcup_{k=N}^{\infty} A_k \\ &\Leftrightarrow \forall N \exists k > N \text{ s.t. } x \in A_k \\ &\Leftrightarrow x \in A_k \text{ for infinitely many } A_k. \end{aligned}$$

Borel-Cantelli lemma // If $A_k \in \mathcal{L}$, $\sum_{k=1}^{\infty} m(A_k) < \infty$, then
 $A_{\infty} := \limsup_{k \rightarrow \infty} A_k \in \mathcal{L}$ and $m(A_{\infty}) = 0$.
 $= \lim_{k \rightarrow \infty} \bigcup_{k=N}^{\infty} A_k$

Def. Let f_n be a sequence of a.e. finite and measurable functions defined on $A \in \mathcal{L}$.

We say f_n converges to f in measure if $\forall \varepsilon > 0$.

$$\boxed{\lim_{k \rightarrow \infty} m(A_k(\varepsilon)) = 0}$$

Prop II If $m(A) < \infty$, then $f_n \rightarrow f$ a.e. $\Rightarrow f_n \rightarrow f$ in measure.

Proof. Recall "downward monotone convergence"

If $\varepsilon > 0$, fixed, we have

$$A > \bigcup_{k=1}^{\infty} A_k(\varepsilon) > \bigcup_{k=2}^{\infty} A_k(\varepsilon) \dots$$

\Rightarrow By downward monotone convergence

$$\lim_{N \rightarrow \infty} m\left(\bigcap_{k=N}^{\infty} A_k(\varepsilon)\right) = m\left(\lim_{N \rightarrow \infty} \bigcap_{k=N}^{\infty} A_k(\varepsilon)\right) = m\left(\bigcap_{k=1}^{\infty} \bigcap_{k=N}^{\infty} A_k(\varepsilon)\right) = 0$$

$$\Rightarrow m(A_N(\varepsilon)) \leq m\left(\bigcup_{k=N}^{\infty} A_k(\varepsilon)\right) \xrightarrow{N \rightarrow \infty} 0$$

$f_n \rightarrow f$ a.e.

$\Rightarrow f_n \rightarrow f$ in measure.

Example Let $E_n = [\frac{k}{2^n}, \frac{k+1}{2^n}] \subset [0, 1]$, where $n = 2^k + k < 2^{k+1}$.

e.g. $E_1 = [0, 1]$, $E_2 = [0, 1]$, $E_3 = [0, \frac{1}{2}]$, $E_4 = [\frac{1}{2}, 1]$, $E_5 = [0, \frac{1}{4}]$, ...

Let $f_n = \chi_{E_n}$.

Then $f_n \rightarrow 0$ in measure [$m(A_n(\varepsilon)) = m(E_n) = \frac{1}{2^n} \rightarrow 0$].

But $f_n \not\rightarrow 0$, $f_n \not\rightarrow 0$ a.e. ($\forall x \in [0, 1]$, $f_n(x) \not\rightarrow 0$)

$$\bigcap_{n=1}^{\infty} \bigcup_{k=N}^{\infty} A_n(\varepsilon) = \bigcap_{n=1}^{\infty} \bigcup_{k=N}^{\infty} E_n = \bigcap_{n=1}^{\infty} [0, 1] = [0, 1]$$

Observe $f_1, f_2, f_3, f_5, f_9, f_{17}, \dots \rightarrow 0$ a.e.

Thm (Riesz). If $f_n \rightarrow f$ in measure $\rightarrow \exists$ Subsequence $\{f_{n_k}\}$ s.t. $f_{n_k} \xrightarrow{\text{a.e.}} f$

Proof. For any $k \in \mathbb{N}$, choose n_k st:

$$m\left(A_{n_k}(\frac{1}{2^k})\right) < \frac{1}{2^k} \quad \forall n \geq n_k.$$

Note we can always arrange s.t. $n_k < n_{k+1} < n_{k+2} \dots$

$$A_{\infty} = \bigcap_{k=1}^{\infty} A_{n_k}(\frac{1}{2^k}).$$

$$\text{Then } \sum_{k=1}^{\infty} m(A_{n_k}(\frac{1}{2^k})) < 1 < \infty.$$

$$\xrightarrow{B-C \text{ lemma}} m(A_{\infty}) = 0.$$

Note $x \notin A_{\infty} \Leftrightarrow x \in A_{n_k}(\frac{1}{2^k})$ for only finitely many k

$$\Leftrightarrow \exists k \text{ s.t. } x \notin A_k(\frac{1}{2^k}) \Rightarrow \forall n \geq k |f_n(x) - f(x)| \leq \frac{1}{2^k}$$

$$f_n(x) \xrightarrow{n \rightarrow \infty} f(x).$$

Uniform convergence \Rightarrow Locally uniformly convergence

(pointwise)

Convergence \Rightarrow a.e. Convergence

Sub-sequence \Rightarrow $m(A) < \infty$

Convergence in measure

outside a set
of measure $< \epsilon$

$\Leftrightarrow \forall \epsilon > 0 \exists N \text{ s.t. } \forall n \geq N \forall A \subset \Omega \text{ s.t. } m(A) < \epsilon$

Recall: $f_n \rightarrow f$ uniformly $\Leftrightarrow \forall \epsilon > 0 \exists N \text{ s.t. } \forall n \geq N \forall x |f_n(x) - f(x)| < \epsilon$

Def: we say $f_n \rightarrow f$ Locally uniformly if for any bounded set E , we have $f_n \rightarrow f$ uniformly

Example $f_n(x) = \frac{x}{n} \quad x \in \mathbb{R}$.

Then $f_n(x) \rightarrow 0$ Locally, pointwise, a.e.

$f_n(x) \not\rightarrow 0$ uniformly

$f_n(x) \not\rightarrow 0$ in measure.

$f_n \rightarrow f$ a.e. $\Leftrightarrow \forall \epsilon > 0 \cdot m(\limsup_{K \rightarrow \infty} A_K(\epsilon)) = 0$.

$f_n \rightarrow f$ in measure $\Leftrightarrow \forall \epsilon > 0 \quad \lim_{K \rightarrow \infty} m(A_K(\epsilon)) = 0$.

Example $f_n(x) = \frac{1}{nx} \chi_{\mathbb{R}^+} = \begin{cases} \frac{1}{nx} & x > 0 \\ 0 & x \leq 0 \end{cases}$

Then $f_n \rightarrow 0$ pointwise, a.e., in measure.

$f_n \not\rightarrow 0$ locally, uniformly.

f_n finite a.e. and measurable.

Thm (Eropob.) If $f_n \rightarrow f$ a.e. then $\forall \epsilon > 0 \exists A_\epsilon \in \mathcal{L}$ with $m(A_\epsilon) < \epsilon$

St. $f_n \rightarrow f$ Locally uniformly on $A \setminus A_\epsilon$.

$\Leftrightarrow \forall x \exists \text{ open set } U \ni x \text{ s.t. } f_n \rightarrow f \text{ uniformly in } U$

Remark: 1) This explains Littlewood's 3rd principle:

- Every convergent seq is "nearly" uniformly convergent

2) We don't require $m(A) < \infty$.

We say a proposition holds locally if $\forall x \exists$ open set $U \ni x$

s.t. the proposition holds in U

Proof: Take A_0 s.t. $m(A_0) = 0$ and

$$f_n \rightarrow f \text{ on } A \setminus A_0.$$

$$\text{Denote } A_{N,m} = \bigcup_{n=N}^{\infty} (A_n \setminus \frac{1}{m}) \cap (A \setminus A_0) \in \mathcal{L}.$$

$$= \{x \mid x \in A \setminus A_0 : \exists n \geq N \text{ s.t. } |f_n(x) - f(x)| > \frac{1}{m}\}.$$

$$\text{Then } \bigcap_{N=1}^{\infty} A_{N,m} = \emptyset. \quad (\text{for a fixed } m).$$

$$(x \in \Leftrightarrow \forall N \geq 1. \exists n \geq N \text{ s.t. } |f_n(x) - f(x)| > \frac{1}{m} \Leftrightarrow f_n(x) \not\rightarrow f(x).)$$

$$\text{Note: } A_{1,m} \supseteq A_{2,m} \dots \supseteq A_{N,m} \supseteq \dots$$

$$\Rightarrow \text{For any } R. \quad (B_{(0,R)} \cap A_{1,m}) \supseteq (B_{(0,R)} \cap A_{2,m}) \supseteq \dots$$

$$\begin{cases} \text{Using fixed } m. & \text{by monotone convergence theorem we get } \lim_{N \rightarrow \infty} m(B_{(0,R)} \cap A_{N,m}) \\ & = 0. \\ \text{In particular, we can find } N_m > 0 \text{ s.t.} & m(B_{(0,R)} \cap A_{N,m}) < \frac{\varepsilon}{2^m}. \end{cases}$$

$$m(B_{(0,R)} \cap A_{N,m}) < \frac{\varepsilon}{2^m}, \quad \forall N \geq N_m.$$

$$\text{Not let } A_\varepsilon = \bigcup_{m=1}^{\infty} (B_{(0,R)} \cap A_{N_m, m}) \setminus A_0 \in \mathcal{L}$$

$$m(A_\varepsilon) < \varepsilon \quad (\text{By sub-additivity})$$

Finally, by construction, we have

$$|f_n(x) - f(x)| < \frac{1}{m}, \quad \forall m \geq 1 \text{ and } \forall x \in (A \setminus A_\varepsilon) \cap B_{(0,R)} \text{ and } n \geq N_m.$$

Now for any bounded set E , with $E \subseteq A \setminus A_\varepsilon$.

we take m large s.t. $E \subseteq B_{(0,R)}$.

Then $|f_n(x) - f(x)| < \frac{1}{m}, \quad \forall x \in E. \quad \underline{n \geq N_m} \Rightarrow f_n \rightarrow f \text{ uniformly on } E$.

2. Approximation by simple functions

Def: A function is simple function if it has the form

$$f(x) = c_1 X_{A_1}(x) + c_2 X_{A_2}(x) + \dots + c_n X_{A_n}(x),$$

where c_1, c_2, \dots, c_n are numbers and $A_1, A_2, \dots, A_n \in \mathcal{L}$

and $A_i \cap A_j = \emptyset$ ($i \neq j$). $f(x) = 0$ if $x \notin \bigcup A_i$. finite

= "Step function": special case of simple functions with each $A_i = \text{box}$.

Note: Any simple function is measurable.

Thm: A function f is measurable if and only if f is the limit of a sequence of simple function $f_n(x)$.

Moreover

- 1) If $f \geq 0$, then we can take $0 \leq f_1 \leq f_2 \dots \rightarrow f$.
- 2) In general, we can take $|f_1| \leq |f_2| \leq \dots \leq |f|$.
- 3) If f is bounded function, we can take $f_n \rightarrow f$ uniformly.

Proof. (a) If $f \geq 0$.

$$\text{Fix } n. \text{ define } f_n(x) = \begin{cases} \frac{k-1}{2^n} & \text{if } \left(\frac{k-1}{2^n} \leq f(x) < \frac{k}{2^n}\right) \quad (k=1, 2, \dots, 2^n), \\ n & \text{if } (f(x) \geq n) \end{cases}$$

→ simple function.

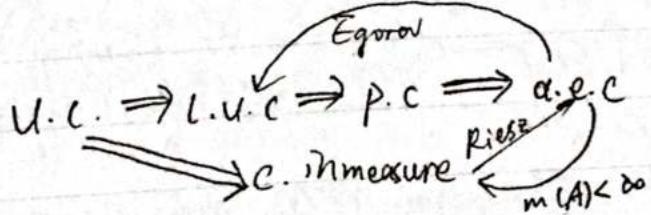
Check: $f_n \leq f_{n+1}$.

- $f_{n+1}(x) - f_n(x) \leq \frac{1}{2^{n+1}}$ if $f < n$
- $f_{n+1}(x) - f_n(x) \leq 1$. if $f \geq n$.

(b) General.

Define f^+ , f^- . $f^+ = \max(f, 0)$, $f^- = \max(-f, 0)$

$$f = f^+ - f^-$$



In all these modes,
the limit f must be measurable

Lee 8

1 Approximation by simple functions

Thm 1: Let $f \geq 0$ be a non-negative measurable function defined on $A \in \mathcal{L}$. Then there exists an increasing sequence of simple functions $0 \leq f_1(x) \leq \dots \leq f_n(x) \leq \dots \leq f(x)$.

Such that $\lim_{n \rightarrow \infty} f_n(x) = f(x)$ moreover, if f is bounded, the $f_n \rightarrow f$ uniformly.

Proof: Define $f_n(x)$ like last Lee.

$$f_n(x) = \begin{cases} \frac{k-1}{2^n} & \text{if } \frac{k-1}{2^n} \leq f(x) < \frac{k}{2^n} \\ n & \text{if } f(x) \geq n. \end{cases} \quad k=1, 2, \dots, n \cdot 2^n$$

Then, since f is measurable, the sets

$$\{x : \frac{k-1}{2^n} \leq f(x) < \frac{k}{2^n}\} \quad \{x : f(x) \geq n\}$$

are measurable. So $f_n(x)$ is a simple function

By def., $f_n(x) \leq f(x)$

We need to compare f_n and f_{n+1} .

$$(1) \text{ If } f(x) < n, \text{ then } \exists k \leq n \cdot 2^n \text{ s.t. } \frac{k-1}{2^n} \leq f(x) < \frac{k}{2^n}.$$

$$\Rightarrow f_n(x) = \frac{k-1}{2^n}$$

$$\Rightarrow f_{n+1}(x) = \frac{2k-2}{2^{n+1}} \text{ or } \frac{2k-1}{2^{n+1}}$$

$$\frac{2k-2}{2^{n+1}} \quad \frac{2k-1}{2^{n+1}} \quad \frac{2k}{2^{n+1}}$$

Thus $f_n(x) \leq f_{n+1}(x)$ and $0 \leq f_{n+1}(x) - f_n(x) \leq \frac{1}{2^{n+1}}$ *

(2) If $f(x) \geq n$, then $f_n(x) = n$.

$$\text{Since } n = n \cdot \frac{2^{n+1}}{2^{n+1}} \leq f(x) \Rightarrow f_{n+1}(x) = \frac{n \cdot 2^{n+1}}{2^{n+1}}, \frac{n \cdot 2^{n+1} + 1}{2^{n+1}}, \dots, \frac{n \cdot 2^{n+1} + 2^{n+1} - 1}{2^{n+1}}$$

$$\Rightarrow 0 \leq f_{n+1}(x) - f_n(x) \leq 1.$$

So we have

$$0 \leq f_1 \leq \dots \leq f_n \leq \dots \leq f$$

and $\lim_{n \rightarrow \infty} f_n(x) = f(x) \Rightarrow$

$$f(x) = \infty \quad \forall n, f_n(x) > n,$$

$f(x) < \infty$. If $f(x) \neq \infty$ then $f_n(x) \rightarrow f(x)$ uniformly.

Moreover, if $f(x) < \infty$, then for $n > M$ only 1/2 occurs.

$$\Rightarrow \forall x, |f_n(x) - f(x)| \leq \frac{1}{2^n} \Rightarrow f_n \rightarrow f \text{ uniformly.}$$

Thm 2 Let f be a measurable function defined on $A \subset \mathbb{Z}$.

There exists a sequence of simple function

$$0 \leq |f_1| \leq |f_2| \leq \dots \leq |f_n| \leq \dots \leq |f|$$

$$\text{s.t. } \lim_{n \rightarrow \infty} f_n(x) = f(x).$$

Moreover, if $|f(x)| < M$, then $f_n \rightarrow f$ uniformly.

Proof: Let $f^+ = \max(f, 0)$, $f^- = \max(-f, 0)$.

then f^+ , f^- are measurable, non-negative and $f = f^+ - f^-$.

Apply I to f^+ , f^- respectively. we get

$$0 \leq g_1 \leq g_2 \leq \dots \leq f^+, \quad 0 \leq h_1 \leq h_2 \leq \dots \leq f^-.$$

$$\begin{matrix} \text{simple} \\ g_n \rightarrow f^+ \end{matrix}$$

$$\begin{matrix} \text{simple} \\ h_n \rightarrow f^- \end{matrix}$$

$$\text{let } f_n = g_n - h_n.$$

$$\rightarrow \text{If } f(x) > 0 \quad |f(x)| = f^+(x). \quad f^-(x) = 0. \rightarrow h_n(x) = 0 \rightarrow |f_n(x)| = |g_n(x)| = g_n(x)$$

$$\rightarrow \text{If } f(x) < 0 \quad \text{Similarly.}$$

$$\leq g_{n+1}(x) = |f_{n+1}(x)|$$

Finally, if $|f(x)| < M$, then $g_n \rightarrow f^+$ uniformly, $h_n \rightarrow f^-$ uniformly

$\Rightarrow f_n \rightarrow f$ uniformly.

Cor: f is measurable $\Leftrightarrow f$ is the limit of a sequence of simple functions.

Cor: If f is measurable, then \exists a sequence of step functions.

f_1, f_2, \dots s.t: $f_n \rightarrow f$ a.e.

Exercise

2 Approximation by continuous functions

Thm (Lusin) Let f be any a.e. finite and measurable function.

defined on $A \subset \mathbb{Z}$. Then $\forall \varepsilon > 0 \exists$ closed set F cat with

$m(A \setminus F) < \varepsilon$ s.t: f is continuous on \bar{F} . for extension

Littlewood's 2nd principle:

"Every measurable function is nearly continuous"

For each n , take closed set F_n with $m(A|F_n) < \frac{\epsilon}{2^n} \cdot \frac{1}{3}$ s.t.

Proof 1: Take a sequence of step functions $f_n \rightarrow f$ a.e. (Cor).
 f_n is continuous on F_n .

By Egorov, $\exists F_\epsilon$ with $m(F_\epsilon) < \frac{\epsilon}{3}$ s.t. $f_n \rightarrow f$ uniformly on $F \setminus F_\epsilon$.

Then $A \setminus ((\bigcup_n A|F_n) \cup F_\epsilon) \in \mathcal{L}$.

Take $F \subset A \setminus ((\bigcup_n A|F_n) \cup F_\epsilon)$ closed s.t. $m((A \setminus ((X|F) \cup F_\epsilon)) \setminus F) < \frac{\epsilon}{3}$

Then $m(A|F) < \epsilon$ and $f_n \rightarrow f$ uniformly on F . f_n is continuous on F .
 $\Rightarrow f$ is continuous on F .

Proof 2: First assume f is simple function, i.e.

$$f = \sum_{i=1}^n c_i \chi_{A_i} \quad A_i \in \mathcal{L}, \quad A_i \cap A_j = \emptyset \text{ and } A = \bigcup_{i=1}^n A_i$$

For each A_i find $F_i \subset A_i$, $m(A_i | F_i) < \frac{\epsilon}{n}$

Then $F = \bigcup_{i=1}^n F_i$ closed, $m(A|F) \leq \sum_{i=1}^n m(A_i | F_i) < \epsilon$.

f is continuous on each F_i . Since it is a constant function

f is continuous on F . Each F_i is closed and $F_i \cap F_j = \emptyset$.

Second assume f is bounded

Then simple functions f_1, f_2, \dots s.t. $f_n \rightarrow f$ uniformly.

By Step 1. $\exists F_n \subset A$ with $m(A|F_n) < \frac{\epsilon}{2^n}$ s.t. f_n is continuous on F_n .

Let $F = \bigcap_{n=1}^{\infty} F_n$. Then F is closed and $m(A|F) < \epsilon$.

Now on F , f_n are continuous $f_n \rightarrow f$ uniformly on F

$\Rightarrow f$ is continuous on F .

Finally, for general case

* Let $g(x) = \frac{f(x)}{1 + |f(x)|} \xrightarrow{[0,1]} \text{one to one}$ By Step 2. $\exists F$ closed with $m(A|F) < \epsilon$
 s.t. g is continuous on F . $f(x) = \frac{g(x)}{1 - g(x)}$ is continuous.

Cor: || f a.e. finite, measurable on $A \in \mathcal{L}$.
 || \exists a continuous function g on \mathbb{R}^n s.t: $m(\{x : f(x) \neq g(x)\}) < \varepsilon$
 → Lusin + Tietze extension thm: || $\forall F \subset \mathbb{R}^n$ closed
 || f continuous on $F \Rightarrow g$ continuous on F
 || s.t: $g = f$ on F
 Moreover, if $|f| \leq M$, then $|g| \leq M$.

Cor: || f a.e. finite measurable
 || Then \exists a sequence of continuous functions f_1, f_2, \dots on \mathbb{R}^n
 || s.t: $f_n \rightarrow f$ a.e.
 By the previous corollary, one can find a sequence of continuous functions f_n s.t: $f_n \rightarrow f$ in measure
 $\Rightarrow \exists$ subseq $f_{n_k} \rightarrow f$ a.e.

Lec 9

Idea: To define $\int f dx$, we first define $\int_A f_n(x) dx$, where f_n is simple. It's simpler if we first let $f_n \rightarrow f$ to be non-negative.

- Consider infinite sum $\sum_{n=1}^{\infty} a_n$

We have two different theories

(1) "Non-negative theory"

If each $a_n > 0$, then $\sum_{n=1}^{\infty} a_n$ always converges (may be $+\infty$)

and we change order $\sum_{n=1}^{\infty} a_n = \lim_{N \rightarrow \infty} \sum_{n=1}^N a_n = \lim_{N \rightarrow \infty} \sum_{n=1}^N a_{\sigma(n)}$

$= \sup \{ \sum_{n \in S} a_n \text{, } S \subset N \text{, } S \text{ is finite.} \}$

$\sigma: N \rightarrow N$ bijective

(2) "absolute convergence theory"

If $\sum_{n=1}^{\infty} |a_n| < +\infty$, then $\sum_{n=1}^{\infty} a_n$ converges, and we can change the order. In particular we have $\sum_{n=1}^{\infty} a_n = \sum_{n=1}^{\infty} a_n^+ - \sum_{n=1}^{\infty} a_n^-$

Similarly, for Lebesgue's integrals, one also have

(1) "Non-negative theory"

→ Only need $f \geq 0$.

→ Generally $\int f dx = \int f^+ - \int f^-$

→ Convergence theorems

| (Lev.) monotone convergence
Faster (Lemma)

(2) "Absolute Convergence Theory"

need $\int_{\mathbb{R}} d|f| dx < \infty$.

| (Lebesgue) dominated convergence theorem

1. Integrals of non-negative simple functions.

• Let $A \in \mathcal{L}$.

for $f(x) = \sum_{i=1}^n a_i \chi_{A_i}(x)$ is non-negative simple function.
 → define $\int_{\mathbb{R}^d} f(x) dx = m(A)$.

• For any non-negative simple function $f(x) = \sum_{n=1}^N (\tilde{a}_n) \chi_{A_n}(x)$.
 we must define $\int_{\mathbb{R}^d} f(x) dx = \sum_{n=1}^N \tilde{a}_n m(A_n)$
 → to ensure linear character.

Remark:

(1) One can check that the definition is independent of representation of f .
 if $\sum_{n=1}^N a_n \chi_{A_n} = \sum_{m=1}^M \tilde{a}_m \chi_{\tilde{A}_m}$ (A_i 's, \tilde{A}_i 's can be overlapped).
 then $\sum_{n=1}^N a_n m(A_n) = \sum_{m=1}^M \tilde{a}_m m(\tilde{A}_m)$

So we always let $A_i \cap A_j = \emptyset$, and a_i is non-zero

(2) If f is defined on $\mathcal{A} \in \mathcal{L}$ we can define $\int_A f dx$ also simple
 $\int_A f dx = \int_{\mathbb{R}^d} f(x) \chi_A(x) dx. \quad \leftarrow \chi_{A_1}(x) \cdot \chi_{A_2}(x) = \chi_{A_1 \cap A_2}(x).$

So we can always extend f to be a simple function on \mathbb{R}^d

s.t. $f = 0$ on $A^c \Rightarrow$ it's enough to handle $\int_{\mathbb{R}^d} f(x) dx$,

and we can assume $\bigcup_{i=1}^n A_i = \mathbb{R}^d$.

Prop: Let $f(x) = \sum_{i=1}^n a_i \chi_{A_i}$ $g(x) = \sum_{j=1}^m b_j \chi_{B_j}$ be non-negative simple functions (assume $A_i \cap A_j = \emptyset$, $\bigcup A_i = \mathbb{R}^d$, $B_i \cap B_j = \emptyset$, $\bigcup B_i = \mathbb{R}^d$).

① For any $c_1, c_2 \geq 0$. $\int_{\mathbb{R}^d} (c_1 f + c_2 g) dx = c_1 \int_{\mathbb{R}^d} f + c_2 \int_{\mathbb{R}^d} g$. Linearity

② If $f \leq g$ $\int_{\mathbb{R}^d} f \leq \int_{\mathbb{R}^d} g$

Monotonicity

③ $\int_{\mathbb{R}^d} f(x) dx = 0 \Leftrightarrow f = 0$ a.e.

④ $\int_{\mathbb{R}^d} f(x) dx < +\infty \Leftrightarrow f$ is a.e. finite. and $m(\text{Supp } f) < +\infty$

$\downarrow x: f(x) \neq 0$

different from
 $c \in \{\exists x: f(x) \neq 0\}$

⑥ $|\int \varphi| \leq \int |\varphi|$. Triangle inequality

⑦ $\int_{E \cup F} \varphi = \int_E \varphi + \int_F \varphi$. E, F are disjoint Additivity

⑧ For any $y \in \mathbb{R}^d$. $\int_{\mathbb{R}^d} f(x+y) dx = \int_{\mathbb{R}^d} f(x) dx$.

Proof. (1) $c_1 f + c_2 g = \sum_{i=1}^n c_1 a_i \chi_{A_i} + \sum_{j=1}^m c_2 b_j \chi_{B_j} = (\sum_{i=1}^n c_1 a_i \chi_{A_i}) \cdot (\sum_{j=1}^m \chi_{B_j})$,
 $+ (\sum_{j=1}^m c_2 b_j \chi_{B_j}) \cdot (\sum_{i=1}^n \chi_{A_i}) = \sum_{i=1}^n \sum_{j=1}^m (c_1 a_i + c_2 b_j) \chi_{A_i \cap B_j}$.
 $= \sum_{i=1}^n \sum_{j=1}^m (c_1 a_i + c_2 b_j) \chi_{A_i \cap B_j}$ $(A_i \cap B_j) \cap (A_{i'} \cap B_{j'}) = \emptyset$
 $\Rightarrow c_1 f + c_2 g \geq 0$. and $(i, j) \neq (i', j')$

$$\begin{aligned} \int_{\mathbb{R}^d} (c_1 f + c_2 g) dx &= \sum_{i=1}^n \sum_{j=1}^m (c_1 a_i + c_2 b_j) m(A_i \cap B_j) \\ &= \sum_{i=1}^n c_1 a_i \left(\sum_{j=1}^m m(A_i \cap B_j) \right) = m(A_i) \\ &+ \sum_{j=1}^m c_2 b_j \left(\sum_{i=1}^n m(A_i \cap B_j) \right) = m(B_j) \\ &= \sum_{i=1}^n c_1 a_i m(A_i) + \sum_{j=1}^m c_2 b_j m(B_j) = c_1 \int_{\mathbb{R}^d} f + c_2 \int_{\mathbb{R}^d} g \end{aligned}$$

(2) Similarly. Let $c_1 = -1$, $c_2 = 1$.

$$g - f = \sum_{i=1}^n \sum_{j=1}^m (b_j - a_i) \chi_{(A_i \cap B_j)}$$
. is a simple function

Since $g \geq f$, $b_j \geq a_i$ if $A_i \cap B_j \neq \emptyset$.

$$\Rightarrow \int_{\mathbb{R}^d} f - g = \sum_{i=1}^n \sum_{j=1}^m (b_j - a_i) m(A_i \cap B_j) \geq 0.$$

$$\Rightarrow \int_{\mathbb{R}^d} f + \int_{\mathbb{R}^d} g - f = \int_{\mathbb{R}^d} g \Rightarrow \int_{\mathbb{R}^d} g \geq \int_{\mathbb{R}^d} f.$$

(3) If $f = 0$ a.e. then if $a_i \neq 0$. then $A_i \subset \{x: f(x) \neq 0\} \Rightarrow m(A_i) = 0$.
 $\Rightarrow \int_{\mathbb{R}^d} f = \sum_{i: a_i \neq 0} a_i \chi_{A_i} = 0$.

If $\int_{\mathbb{R}^d} f dx = \sum_{i=1}^n a_i m(A_i) = 0$. then $\forall i$ $a_i m(A_i) = 0$.
 $\Rightarrow \{x: f(x) \neq 0\} = \bigcup_{a_i \neq 0} \{x: f(x) = a_i\} = \bigcup_{a_i \neq 0} A_i$.

$$\Rightarrow m(\{x: f(x) \neq 0\}) = 0.$$

(4) If f is finite a.e. and $m(\text{supp } f) < \infty$

Then let $a = \sup \{a_i : a_i < +\infty\} < +\infty$

we have

$$\int_{\mathbb{R}^d} f(x) dx = \sum_{i=1}^n a_i m(A_i) = \sum_{\substack{0 < a_i < a \\ 0 < a_i < +\infty}} a_i m(A_i) \leq a \sum_{\substack{0 < a_i < a \\ 0 < a_i < +\infty}} m(A_i)$$

$$= a \cdot m(\bigcup A_i) \leq a \cdot m(\text{supp } f) < +\infty.$$

If $m(\{x : f(x) = +\infty\}) > 0$, then $\int_{\mathbb{R}^d} f(x) dx = \sum_{i=1}^n a_i m(A_i) \geq +\infty \cdot m(\{x : f(x) = +\infty\}) = +\infty$.

If $m(\text{supp } f) = +\infty$ let $\tilde{a} = \inf \{a_i : a_i > 0\} > 0$.

Then $\int_{\mathbb{R}^d} f(x) dx = \sum_{i=1}^n a_i m(A_i) \geq \sum_{a_i > 0} \tilde{a} m(A_i) = \tilde{a} \cdot m(\text{supp } f) = +\infty$.

(5) As a function of X

$$f(x+y) = \sum_{i=1}^n a_i \chi_{A_i-y}(x). \quad A_i-y = \{x-y : x \in A_i\}.$$

By translation invariance of Lebesgue measure, we have $A_i-y \in \mathcal{L}$ and $m(A_i-y) = m(A_i)$.

$\Rightarrow f \geq 0$ and is a simple function and

$$\int_{\mathbb{R}^d} f(x+y) dx = \sum_{i=1}^n a_i m(A_i-y) = \sum_{i=1}^n a_i m(A_i) = \int_{\mathbb{R}^d} f(x) dx.$$

2. Integration of non-negative measurable function

Now let f be a non-negative measurable function

want: define $\int_{\mathbb{R}^d} f(x) dx = \lim_{n \rightarrow \infty} \int_{\mathbb{R}^d} f_n(x) dx$ where $f_n \uparrow f$ f_n simple.

Then: need to check it's well-defined \Rightarrow complicated.

Instead we define (for measurable function).

$$\int_{\mathbb{R}^d} f(x) dx = \sup \left\{ \int_{\mathbb{R}^d} h(x) dx : h \text{ is non-negative simple and } h \leq f \right\}.$$

This is always well defined.

Remark: If f is defined on $A \in \mathcal{L}$. then

$$\int_A f(x) dx = \int_{\mathbb{R}^d} f(x) \chi_A(x) dx.$$

So. it's enough to handle $\int_{\mathbb{R}^d} f(x) dx$.

(2) In the case f is non-measurable ($\text{st} \delta \geq 0$)

one can define lower Lebesgue integral $\int_{\mathbb{R}^d} f(x) dx := \sup \left\{ \int_{\mathbb{R}^d} h(x) dx : h \geq 0, \text{ simple and } h \leq f \right\}$. ($\underline{\int}_{\mathbb{R}^d} f(x) dx$)
 and upper Lebesgue integral $\int_{\mathbb{R}^d} f(x) dx := \inf \left\{ \int_{\mathbb{R}^d} h(x) dx : h \geq 0, \text{ simple and } h \geq f \right\}$. ($\overline{\int}_{\mathbb{R}^d} f(x) dx$)

$$\underline{\int}_{\mathbb{R}^d} = \underline{\int}_{\mathbb{R}^d}$$

One can prove: if f is measurable and bounded and $m(\sup f) < \infty$
 then $\underline{\int}_{\mathbb{R}^d} f dx = \overline{\int}_{\mathbb{R}^d} f dx$.

Prop: Let $f(x), g(x)$ be non-negative measurable functions

$$\textcircled{1} \quad c_1, c_2 \geq 0. \quad \int_{\mathbb{R}^d} (c_1 f + c_2 g) dx = c_1 \int_{\mathbb{R}^d} f + c_2 \int_{\mathbb{R}^d} g$$

$$\textcircled{2} \quad \text{If } f \leq g \text{ then } \int_{\mathbb{R}^d} f dx \leq \int_{\mathbb{R}^d} g dx.$$

$$\textcircled{3} \quad \int_{\mathbb{R}^d} f(x) dx = 0 \Leftrightarrow f = 0 \text{ a.e.}$$

$$\textcircled{4} \quad \text{[Markov's inequality]} \quad \text{For any } 0 < \lambda < \infty, \text{ one has } m(\{x \in \mathbb{R}^d : f(x) \geq \lambda\}) \leq \frac{1}{\lambda} \int_{\mathbb{R}^d} f(x) dx.$$

$$\textcircled{5} \quad \text{For any } y \in \mathbb{R}^d \quad \int_{\mathbb{R}^d} f(x+y) dx = \int_{\mathbb{R}^d} f(x) dx.$$

Proof (1) We first prove $\forall c > 0. \quad \int_{\mathbb{R}^d} c f dx = c \int_{\mathbb{R}^d} f dx$.

This is true because

$$\begin{aligned} \int_{\mathbb{R}^d} c f(x) dx &= \sup \left\{ \int_{\mathbb{R}^d} h(x) dx : h \geq 0, \text{ simple, } h \leq c f \right\} \\ &= \sup \left\{ c \int_{\mathbb{R}^d} \tilde{h}(x) dx : \tilde{h} \geq 0, \text{ simple, } \tilde{h} \leq f \right\} \quad \tilde{h} = \frac{h}{c}. \\ &= c \sup \left\{ \int_{\mathbb{R}^d} \tilde{h}(x) dx : \tilde{h} \geq 0, \text{ simple, } \tilde{h} \leq f \right\} \\ &= c \int_{\mathbb{R}^d} f(x) dx. \end{aligned}$$

It remains to prove $\int_{\mathbb{R}^d} (f+g) dx = \int_{\mathbb{R}^d} f dx + \int_{\mathbb{R}^d} g dx$.

[We will postpone this to next time].

12) If $0 \leq h \leq f$ then $0 \leq h \leq g$.

$$\int_{\mathbb{R}^d} f(x) dx = \sup \left\{ \int_{\mathbb{R}^d} h(x) dx \mid h \geq 0, \text{ simple}, h \leq f \right\}.$$

$$\leq \sup \left\{ \int_{\mathbb{R}^d} h(x) dx \mid h \geq 0, \text{ simple}, h \leq g \right\} = \int_{\mathbb{R}^d} g(x) dx$$

(4) Let $g(x) = \lambda \chi_{\{x: f(x) \geq \lambda\}}$. Then $g \geq 0$. simple.

$$g(x) \leq f(x) \quad \forall x.$$

$$\underline{\text{By 12)}} \quad \int_{\mathbb{R}^d} g(x) dx \leq \int_{\mathbb{R}^d} f(x) dx.$$

(definition of $\int_{\mathbb{R}^d} f(x) dx$). $\lambda \cdot m(\{x: f(x) \geq \lambda\})$)

(3) If $f = 0$ a.e. and $0 \leq h \leq f$ then $h = 0$ a.e. $\int_{\mathbb{R}^d} h(x) dx = 0$

$$\Rightarrow \int_{\mathbb{R}^d} f(x) dx = \sup \left\{ \int_{\mathbb{R}^d} h(x) dx \mid 0 \leq h \leq f, h \text{ simple} \right\} = 0.$$

Conversely. Suppose $\int_{\mathbb{R}^d} f(x) dx = 0$.

$$\begin{aligned} \text{Then } m(\{x: f(x) \neq 0\}) &= m\left(\bigcup_{k=1}^{\infty} \{x: f(x) > \frac{1}{k}\}\right) \leq \sum_{k=1}^{\infty} m(\{x: f(x) > \frac{1}{k}\}) \\ &\leq \sum_{k=1}^{\infty} k \int_{\mathbb{R}^d} f(x) dx = 0. \Rightarrow f = 0 \text{ a.e.} \end{aligned}$$

(5) If h is simple $0 \leq h \leq f$.

then as a function of x . $h(x+y)$ is simple and

$$0 \leq h(x+y) \leq f(x+y) \quad \text{in fact.}$$

$$\Rightarrow \int_{\mathbb{R}^d} f(x+y) dx \geq \sup \left\{ \int_{\mathbb{R}^d} h(x+y) dx \mid h \text{ simple}, 0 \leq h \leq f \right\}.$$

$$= \sup \left\{ \int_{\mathbb{R}^d} h(x) dx \mid h \text{ simple}, 0 \leq h \leq f \right\} = \int_{\mathbb{R}^d} f(x) dx.$$

Conversely, if $0 \leq h(x+y) \leq f(x+y)$ the $h(x)$ simple and $0 \leq h(x) \leq f(x)$.
Similarly.

Def: We say a non-negative measurable f is integrable

$$\int_{\mathbb{R}^d} f(x) dx < +\infty$$

We have seen

A non-negative simple function f is integrable

$\Leftrightarrow f$ is finite a.e. and $m(\text{Supp } f) < +\infty$.

This is no longer true for measurable functions \nRightarrow

e.g. $f(x) = \sum_{n=1}^{\infty} \frac{1}{n^2} \chi_{[n, n+1]}$. \rightarrow limit of a series of simple functions

$$\int_R d f(x) dx = \int_R d \left(\sum_{n=1}^{\infty} \frac{1}{n^2} \chi_{[n, n+1]} \right) dx$$

$$= \lim_{N \rightarrow \infty} \int_R d \left(\sum_{n=1}^{N+1} \frac{1}{n^2} \chi_{[n, n+1]} \right) dx = \lim_{N \rightarrow \infty} \sum_{n=1}^{N+1} \frac{1}{n^2} = \frac{\pi^2}{6} < \infty. \text{ Integrable.}$$

But. $\text{Supp } f = +\infty$.

Prop II Suppose f is integrable \Rightarrow then f is a.e. finite.

Proof: $m(\{x: f(x)=+\infty\}) = m(\bigcap_{k=1}^{\infty} \{x: f(x) > k\}) \leq m(\{x: f(x) > k\}) \quad \forall k > 0$
 $\leq \frac{1}{k} \int_R d f(x) dx$

Let $k \rightarrow \infty$. we get $m(\{x: f(x)=+\infty\}) = 0$.

Lee 10

Today: Limits of integrals

Recall (Lee. 4) Monotonicity of measure

$$A_1 \subset A_2 \subset \dots \quad A_i \in \mathcal{L} \Rightarrow \lim_{n \rightarrow \infty} A_n = \bigcup_{n=1}^{\infty} A_n \in \mathcal{L}$$

$$\text{and } m(\lim_{n \rightarrow \infty} A_n) = \lim_{n \rightarrow \infty} m(A_n)$$

Interpret v.i.e. "integrals"

$\chi_{A_1} \leq \chi_{A_2} \leq \dots \Rightarrow \lim_{n \rightarrow \infty} \chi_{A_n}$ is simple, and

$$\int_R d \underbrace{\lim_{n \rightarrow \infty} \chi_{A_n}}_{\text{con } A_n} dx = \lim_{n \rightarrow \infty} \int_R d A_n dx \Leftrightarrow \int_R d x = \lim_{n \rightarrow \infty} \int_{A_n} d x.$$

Levi's monotone convergence theorem

In $0 \leq f_1 \leq f_2 \leq f_3 \leq \dots$, \exists a monotone increasing sequence of non-negative measurable functions

then $\int \lim_{n \rightarrow \infty} f_n$ is a non-negative measurable and

$$\text{II } \int_{R^d} f_n dx = \lim_{n \rightarrow \infty} \int_{R^d} f_n(x) dx.$$

Proof: Let's first prove a simple case.

Lemma: Suppose $A_1 \subset A_2 \subset \dots$ and $h(x) = \sum_{i=1}^m a_i \chi_{B_i}$ is simple.
 $\| h \geq 0 \text{. Then } \lim_{n \rightarrow \infty} \int_{A_n} h dx = \int_{\lim_{n \rightarrow \infty} A_n} h dx$

(let $f_n = h \cdot \chi_{A_n}$. Then $f_n \geq 0$, $f_n \uparrow$).

$$\begin{aligned} \text{Proof. } \int_{A_n} h dx &= \sum_{i=1}^m a_i \int_{A_n} \chi_{B_i} dx = \sum_{i=1}^m a_i m(A_n \cap B_i) \\ &\rightarrow \sum_{i=1}^m a_i m(\lim_{n \rightarrow \infty} A_n \cap B_i) \\ &= \sum_{i=1}^m a_i \int_{\lim_{n \rightarrow \infty} A_n} \chi_{B_i} dx = \int_{\lim_{n \rightarrow \infty} A_n} h dx. \end{aligned}$$

Proof of Levi's

$$f_n \leq \lim_{n \rightarrow \infty} f_n \Rightarrow \int_{R^d} f_n dx \leq \int_{R^d} \lim_{n \rightarrow \infty} f_n dx \Rightarrow \lim_{n \rightarrow \infty} \int_{R^d} f_n dx \leq \int_{R^d} (\lim_{n \rightarrow \infty} f_n) dx$$

Take any non-negative simple function, $0 \leq h \leq \lim_{n \rightarrow \infty} f_n$, Fixed ε .

Let $A_n = \{x : f_n(x) \geq (1-\varepsilon)h(x)\}$.

Then • $A_n \subset A_{n+1}$

$$\bullet \bigcup_{n=1}^{\infty} A_n = \bigcup_{n=1}^{\infty} A_n = R^d.$$

$$\text{By Lemma. } \int_{A_n} h(x) dx \rightarrow \lim_{n \rightarrow \infty} \int_{A_n} h(x) dx = \int_{R^d} h(x) dx.$$

$$\Rightarrow \int_{R^d} f_n(x) dx \geq \int_{A_n} f_n(x) dx \geq (1-\varepsilon) \int_{A_n} h(x) dx$$

$$\Rightarrow \lim_{n \rightarrow \infty} \int_{R^d} f_n(x) dx \geq (1-\varepsilon) \lim_{n \rightarrow \infty} \int_{A_n} h(x) dx = (1-\varepsilon) \int_{R^d} h(x) dx.$$

$$\Rightarrow \lim_{n \rightarrow \infty} \int_{R^d} f_n(x) dx \geq (1-\varepsilon) \sup \left\{ \int_{R^d} h(x) dx \mid 0 \leq h \leq \lim_{n \rightarrow \infty} f_n, h \text{ simple} \right\} \int_{R^d} h(x) dx$$

$$= (1-\varepsilon) \int_{R^d} (\lim_{n \rightarrow \infty} f_n) dx$$

$$\text{Let } \varepsilon \rightarrow 0. \lim_{n \rightarrow \infty} \int_{R^d} f_n(x) dx \geq \int_{R^d} \lim_{n \rightarrow \infty} f_n(x) dx$$

DURSTORY BEGINS
another def of $\int_R d f dx$.

Cor II If $0 \leq f_1 \leq f_2 \leq \dots \rightarrow f$ f_n is simple. $\int_R d f dx = \lim_{n \rightarrow \infty} \int_R d f_n dx$.

Cor II $\int_R d (f+g) dx = \int_R d f dx + \int_R d g dx$.

Proof: Take simple function $f_n \nearrow f$ $g_n \nearrow g$.

$$\Rightarrow f_n + g_n \nearrow f+g$$

$$\text{Then } \int_R d (f+g) dx = \lim_{n \rightarrow \infty} \int_R d (f_n + g_n) dx = \lim_{n \rightarrow \infty} \int_R d f_n dx + \lim_{n \rightarrow \infty} \int_R d g_n dx \\ = \int_R d f dx + \int_R d g dx$$

Cor II $\int_R d (\sum_{i=1}^n f_i) dx = \sum_{i=1}^n \int_R d f_i dx$ where f_i are non-negative

Cor (Integration by terms) // let f_1, \dots, f_n, \dots be nonnegative measurable. Then $\int_R d (\sum_{i=1}^{\infty} f_i) dx$
 $= \sum_{i=1}^{\infty} \int_R d f_i dx$.

Proof: Let $g_N = \sum_{i=1}^N f_i$ then $g_1 \leq g_2 \leq g_3 \dots$

$$\Rightarrow \int_R d \sum_{i=1}^{\infty} f_i dx = \int_R d \lim_{N \rightarrow \infty} g_N dx = \lim_{N \rightarrow \infty} \int_R d g_N dx = \lim_{N \rightarrow \infty} \sum_{i=1}^N \int_R d f_i dx \\ = \sum_{i=1}^{\infty} \int_R d f_i dx$$

Cor: Suppose $A_i \in \mathcal{L}$. $A_i \cap A_j = \emptyset$. Then for any non-negative measurable function f $\int_{\bigcup_{i=1}^{\infty} A_i} f dx = \sum_{i=1}^{\infty} \int_{A_i} f dx$.

Proof: (let $g_n = f(x) \cdot \chi_{\bigcup_{i=1}^n A_i}$ Then $g_n \leq g_{n+1}$)

$$\Rightarrow \int_{\bigcup_{i=1}^{\infty} A_i} f dx = \int_{\bigcup_{i=1}^{\infty} A_i} g_n dx = \lim_{n \rightarrow \infty} \int_R d g_n dx = \rightarrow A_i \cap A_j = \emptyset$$

$$\lim_{n \rightarrow \infty} \int_R d f(x) \cdot \chi_{\bigcup_{i=1}^n A_i} = \left(\lim_{n \rightarrow \infty} \int_R d f(x) \right) \sum_{i=1}^n \chi_{A_i}(x) dx \\ = \lim_{n \rightarrow \infty} \sum_{i=1}^n \int_R d f(x) \chi_{A_i}(x) dx = \sum_{i=1}^{\infty} \int_{A_i} f dx.$$

Remark: "Abstract measure"

$M: \mathcal{F} \rightarrow [0, \infty]$, s.t: $M(\emptyset) = 0$

$$M(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} M(A_i)$$

Countable additivity

We can define $M_f: \mathcal{L} \rightarrow [0, +\infty]$.

$$\text{s.t. } M_f(A) = \int_A f dx \quad (\forall f, f \geq 0 \text{ and measurable})$$

$$\text{Then } M_f(\emptyset) = \int_{\emptyset} f dx = 0.$$

$$M_f\left(\bigcup_{i=1}^{\infty} A_i\right) = \int_{\bigcup_{i=1}^{\infty} A_i} f dx = \sum_{i=1}^{\infty} \int_{A_i} f dx = \sum_{i=1}^{\infty} M_f(A_i).$$

Cor Let $f \geq 0$ be an a.e. finite and measurable on $A \in \mathcal{L}$

$$\text{and } m(A) < +\infty. \text{ Then } \int_A f dx = \int_0^{\infty} m(\{x : f(x) \geq t\}) dt$$

$$= \lim_{\delta \rightarrow 0} \sum_{k=0}^{\infty} (t_{k+1} - t_k) m(\{x : f(x) \geq t_k\}).$$

↑ Riemann
↓ Severe
where $|t_{k+1} - t_k| \leq \delta$

Proof: By def $A_k \supset A_{k+1} \supset \dots$

$$t_k m(A_k \setminus A_{k+1}) \leq \int_{A_k \setminus A_{k+1}} f(x) dx \leq t_{k+1} m(A_k \setminus A_{k+1})$$

$$\Rightarrow \sum_{k=0}^{\infty} t_k (m(A_k) - m(A_{k+1})) \leq \int_A f dx \leq \sum_{k=0}^{\infty} t_{k+1} (m(A_k) - m(A_{k+1}))$$

$$\text{from } \sum_{k=0}^{\infty} t_k (m(A_k) - m(A_{k+1})) = \sum_{k=0}^{\infty} t_{k+1} m(A_k) - t_k m(A_k)$$

$$= \sum_{k=0}^{\infty} (t_{k+1} - t_k) m(A_{k+1})$$

$$= \sum_{k=0}^{\infty} (t_{k+1} - t_k) (m(A_{k+1}) - m(A_k))$$

$$+ \sum_{k=0}^{\infty} (t_{k+1} - t_k) m(A_k)$$

$$\leq \delta \sum_{k=0}^{\infty} m(A_{k+1}) m(A_k)$$

$$= \delta m(A) \rightarrow 0 \text{ as } \delta \rightarrow 0$$

$m(A) < +\infty$

Fatou Lemma // Suppose f_n is a seq, $f_n \geq 0$ measurable
// Then $\int_{\mathbb{R}^n} \liminf_{n \rightarrow \infty} f_n dx \leq \liminf_{n \rightarrow \infty} \int_{\mathbb{R}^n} f_n dx$.

Proof: For $i \geq k$. We have $\inf_{n \geq k} f_n \leq f_i$.

$$\rightarrow \int_{\mathbb{R}^n} \inf_{n \geq k} f_n dx \leq \inf_{j \geq k} \int_{\mathbb{R}^n} f_j dx.$$

$$\int_{\mathbb{R}^n} g_K dx \downarrow K \rightarrow \infty$$

$$\text{Let } g_K = \inf_{n \geq k} f_n \leq g_{K+1} \leq \dots$$

$$\liminf_{n \rightarrow \infty} \int_{\mathbb{R}^n} f_n dx \leq \liminf_{K \rightarrow \infty} \int_{\mathbb{R}^n} f_K dx$$

$$f_n = \begin{cases} n & 0 < x < \frac{1}{n} \\ 0 & \text{else} \end{cases}$$

Pset 3 - Part 1

2. Lemma 1 $L: \mathbb{R}^d \rightarrow \mathbb{R}^d$ L : Linear transformation

Then $m^*(L(A)) = |\det L| \cdot m^*(A)$.

$\forall A \in \mathcal{L}$

Pf:

① box $\bigcup_{i=1}^n B_i$ open set $\bigcup_{i=1}^n B_i \in \mathcal{L}$ ② $A \in \mathcal{L}$

$$\begin{aligned} \textcircled{2} \quad m^*(L(A)) &= m^*(L(\bigcup_{i=1}^n B_i)) = m^*(\bigcup_{i=1}^n L(B_i)) = \sum m^*(L(B_i)) = |\det L| \geq m^*(B_i) \\ &= |\det L| \cdot m^*(A). \end{aligned}$$

Pset 3 - Part 2

4. f measurable find $c \in \mathbb{R}$ $f^{-1}(c) \notin \mathcal{L}$.

Pf: 若 f 高 $\exists B \subset A$ ($m(B) > 0$) $m(f(B)) = 0$. \exists 不可测 $w \in B$.
可逆 $0 \leq m(f(w)) \leq m(f(B)) = 0$. \therefore choose $f(w)$ as c .

$$\begin{array}{ccc} B & \xrightarrow{f} & f(B) \\ w \in B & \xrightarrow{f} & f(w) \in f(B) \end{array}$$

Claim: $\exists w \in f^{-1}(c)$ w 不可测, 已测集有不可测子集
 $m(f(w)) > 0$.

不然 $m(A) > 0$, $A \subset [0,1]$.

$\exists w = A/\mathbb{Q}$: (定义~关系) $x \sim y: x-y \in \mathbb{Q}$. 在每个等价类中
取一个元素相交 w . $\exists w_r = w+r$, $r \in \mathbb{Q} \cap [-1,1]$.

$\Rightarrow A \subset \bigcup_{r \in \mathbb{Q} \cap [-1,1]} w_r + A$ $w \in \mathcal{L}$. $0 < m(A) \leq \sum m(w_r) \leq 3$ 矛盾.
 $\rightarrow w$ 不可测.

Lec 11

1 Definition of Lebesgue integral for general measurable functions
 (Absolute integrability)

- Let f be a measurable function on \mathbb{R}^d

(It's enough to assume f is a.e measurable).

$$f(x) = f^+(x) - f^-(x), \quad |f(x)| = f^+(x) + f^-(x)$$

\Rightarrow both f^+ and f^- are non-negative measurable functions

Define: $\int_{\mathbb{R}^d} f dx = \int_{\mathbb{R}^d} f^+ dx - \int_{\mathbb{R}^d} f^- dx$.

(1) If at least one of $\int_{\mathbb{R}^d} f^+ dx$, $\int_{\mathbb{R}^d} f^- dx$ is finite, then we can define

$$\int_{\mathbb{R}^d} f dx = \int_{\mathbb{R}^d} f^+ dx - \int_{\mathbb{R}^d} f^- dx.$$

(2) we say f is integrable if

$\int_{\mathbb{R}^d} f^+$ and $\int_{\mathbb{R}^d} f^-$ are finite
 $(\Rightarrow \int_{\mathbb{R}^d} f(x) dx$ is finite)

Notation: $L'(\mathbb{R}^d)$ = the set of all integrable functions on \mathbb{R}^d .

Remark: • For any measurable set A and a measurable function f on A , we can define

$$\int_A f = \int_{\mathbb{R}^d} f \cdot \chi_A \rightarrow \begin{cases} f(x) & x \in A \\ 0 & x \notin A \end{cases}$$

zero extension of f .

- If $f = g$ a.e. on A , then $\int_A f dx = \int_A g dx$.
- Notation $L'(A)$ = integrable functions on A .

Prop: $f \in L^1(\mathbb{R}^d) \iff |f| \in L^1(\mathbb{R}^d) (\int_{\mathbb{R}^d} |f| dx < +\infty)$.

Proof: $\cdot f \in L^1(\mathbb{R}^d) \Rightarrow \int_{\mathbb{R}^d} f^+ < \infty, \int_{\mathbb{R}^d} f^- < \infty \Rightarrow \int_{\mathbb{R}^d} |f| dx = \int_{\mathbb{R}^d} f^+ + \int_{\mathbb{R}^d} f^- < \infty$

$\cdot \int_{\mathbb{R}^d} |f| dx < \infty \Rightarrow \int_{\mathbb{R}^d} f^+ \leq \int_{\mathbb{R}^d} |f| < +\infty \Rightarrow f \in L^1(\mathbb{R}^d)$.

Cor: $f \in L^1(\mathbb{R}^d) \Rightarrow f$ is a.e. finite.

Proof: $\int_{\mathbb{R}^d} |f| < +\infty \Rightarrow |f|$ is a.e finite. $\Rightarrow f$ is a.e finite.

Cor: $L^1(\mathbb{R}^d)$ is a linear space.

Proof: Suppose $f_1, f_2 \in L^1(\mathbb{R}^d), c_1, c_2 \in \mathbb{R}$. Then

$$\begin{aligned} \int_{\mathbb{R}^d} |c_1 f_1 + c_2 f_2| dx &\leq \int_{\mathbb{R}^d} (|c_1| |f_1| + |c_2| |f_2|) dx \\ &= |c_1| \int_{\mathbb{R}^d} |f_1| dx + |c_2| \int_{\mathbb{R}^d} |f_2| dx < \infty. \\ \Rightarrow c_1 f_1 + c_2 f_2 &\in L^1(\mathbb{R}^d). \end{aligned}$$

Def: Let $f \in L^1(\mathbb{R}^d)$ We call the quantity

$\|f\|_{L^1(\mathbb{R}^d)} := \int_{\mathbb{R}^d} |f| dx < \infty$.
the L^1 -norm of f .

It's obvious that

$$\textcircled{1} \|cf\|_{L^1(\mathbb{R}^d)} = |c| \|f\|_{L^1(\mathbb{R}^d)}$$

$$\textcircled{2} \|f_1 + f_2\|_{L^1(\mathbb{R}^d)} \leq \|f_1\|_{L^1(\mathbb{R}^d)} + \|f_2\|_{L^1(\mathbb{R}^d)}$$

$$\textcircled{3} \|f\|_{L^1(\mathbb{R}^d)} = 0 \iff f = 0 \text{ a.e.}$$

Given $f, g \in L^1(\mathbb{R}^d)$ we can define L^1 -distance.

$$d_{L^1(\mathbb{R}^d)}(f, g) = \|f - g\|_{L^1(\mathbb{R}^d)}.$$

One can prove

$(\bar{L}^1(\mathbb{R}^d), d_{L^1})$ is complete metric space

For $p \geq 1$ $\|f\|_{L^p(\mathbb{R}^d)} = \left(\int_{\mathbb{R}^d} |f|^p dx \right)^{1/p}$
 $L^p(\mathbb{R}^d) = \{ f : \|f\|_{L^p} < \infty \}$

2. Properties of Lebesgue integrals (Absolute integral).

Prop. Let $f_1, f_2 \in L^1(\mathbb{R}^d)$. Then

- (1) (linearity) $\int_{\mathbb{R}^d} (c_1 f_1 + c_2 f_2) dx = c_1 \int_{\mathbb{R}^d} f_1 dx + c_2 \int_{\mathbb{R}^d} f_2 dx \quad \forall c_1, c_2 \in \mathbb{R}$
- (2) (monotonicity) If $f_1 \leq f_2$ a.e. then $\int_{\mathbb{R}^d} f_1 \leq \int_{\mathbb{R}^d} f_2$.
- (3) (translation invariance) $\int_{\mathbb{R}^d} f(x+y) dx = \int_{\mathbb{R}^d} f(x) dx$.

Proof: (1) It is enough to prove.

$$(a) \int_{\mathbb{R}^d} cf dx = c \int_{\mathbb{R}^d} f dx \quad (c \in \mathbb{R})$$

$$(b) \int_{\mathbb{R}^d} f_1 + f_2 dx = \int_{\mathbb{R}^d} f_1 dx + \int_{\mathbb{R}^d} f_2 dx$$

To prove (a)

If $c \geq 0$ then $(cf)^+ = cf^+$, $(cf)^- = cf^-$

$$\Rightarrow \int_{\mathbb{R}^d} cf dx = \int_{\mathbb{R}^d} (cf)^+ dx - \int_{\mathbb{R}^d} (cf)^- dx$$

$$= c \int_{\mathbb{R}^d} f^+ dx - c \int_{\mathbb{R}^d} f^- dx = c \left(\int_{\mathbb{R}^d} f^+ dx - \int_{\mathbb{R}^d} f^- dx \right) = c \int_{\mathbb{R}^d} f dx$$

If $c < 0$, $(cf)^+ = (-c) \cdot f^+$, $(cf)^- = (-c) \cdot f^-$

$$\Rightarrow \int_{\mathbb{R}^d} cf dx = \int_{\mathbb{R}^d} (-c) \cdot f^+ dx - \int_{\mathbb{R}^d} (-c) \cdot f^- dx$$

$$= (-c) \left(\int_{\mathbb{R}^d} f^+ dx - \int_{\mathbb{R}^d} f^- dx \right) = c \left(\int_{\mathbb{R}^d} f^+ dx - \int_{\mathbb{R}^d} f^- dx \right) = c \int_{\mathbb{R}^d} f dx$$

To prove (b), we denote $g = f_1 + f_2$. Then

$$g^+ - g^- = g = f_1 + f_2 = (f_1^+ - f_1^-) + (f_2^+ - f_2^-)$$

$$\Rightarrow g^+ + f_1^- + f_2^- = g^- + f_1^+ + f_2^+$$

$$\Rightarrow \int_{\mathbb{R}^d} g^+ + \int_{\mathbb{R}^d} f_1^- + \int_{\mathbb{R}^d} f_2^- = \int_{\mathbb{R}^d} g^- + \int_{\mathbb{R}^d} f_1^+ + \int_{\mathbb{R}^d} f_2^+$$

(each term is $< \infty$, thus can change order)

$$\Rightarrow \int_{\mathbb{R}^d} (f_1 + f_2) dx = \int_{\mathbb{R}^d} g^+ dx - \int_{\mathbb{R}^d} g^- dx$$

$$= (\int_{\mathbb{R}^d} f_1^+ dx - \int_{\mathbb{R}^d} f_1^- dx) + (\int_{\mathbb{R}^d} f_2^+ dx - \int_{\mathbb{R}^d} f_2^- dx)$$

$$= \int_{\mathbb{R}^d} f_1 dx + \int_{\mathbb{R}^d} f_2 dx$$

(2) If $f_1 \leq f_2$ a.e. then $\int_{\mathbb{R}^d} f_1 dx \leq \int_{\mathbb{R}^d} f_1 + \int_{\mathbb{R}^d} (f_2 - f_1) \stackrel{\text{linearity}}{\geq 0} = \int_{\mathbb{R}^d} f_2 dx.$

(3) $\int_{\mathbb{R}^d} f(x+y) dx = \int_{\mathbb{R}^d} f^+(x+y) dx - \int_{\mathbb{R}^d} f^-(x+y) dx$
 $= \int_{\mathbb{R}^d} f^+(x) dx - \int_{\mathbb{R}^d} f^-(x) dx$
 $= \int_{\mathbb{R}^d} f(x) dx.$

$\int_A f(x+y) dx = \int_{A+3y} f(x) dx.$

Prop (Countable additivity w.r.t. domain) \rightarrow "signed measure"

Let $A_i \cap A_j = \emptyset$. $A = \bigcup_{i=1}^{\infty} A_i$ Suppose $f \in L^1(A)$.

Then $\int_A f dx = \sum_{i=1}^{\infty} \int_{A_i} f dx$ $\left[\begin{array}{l} f \in L^1(A_i) \\ \forall i \end{array} \right]$

Proof: $f \in L^1(A) \Rightarrow \int_A |f| < \infty \Rightarrow \int_A \chi_{A_i} |f| = \int_A |f| \chi_{A_i} < \int_A |f| < \infty$
 $\Rightarrow f \in L^1(A_i)$.

According to the "countable additivity w.r.t. domain"

In non-negative theory:

$$\int_A f^+ dx = \sum_{i=1}^{\infty} \int_{A_i} f^+ dx$$

$$\int_A f^- dx = \sum_{i=1}^{\infty} \int_{A_i} f^- dx$$

$$\Rightarrow \int_A f dx = \int_A (f^+ - f^-) dx = \sum (\int_{A_i} f^+ - \int_{A_i} f^-) = \sum_{i=1}^{\infty} \int_{A_i} f dx.$$

In non-negative theory $\int_{\mathbb{R}^d} f dx = 0 \Leftrightarrow f = 0$ a.e.

This is no longer true for general $f \in L^1(\mathbb{R}^d)$

Prop: Let $A \in \mathcal{L}$. $f \in L^1(A)$ Then

$$f = 0 \text{ a.e. on } A \Leftrightarrow \forall B \subset A \quad \int_B f dx = 0.$$

Proof. \Rightarrow Obviously ($f^+ = 0$ a.e. $f^- = 0$ a.e.).

$$\Leftarrow \text{Let } B_1 = \{x : f(x) > 0\}, \quad B_2 = \{x : f(x) < 0\}.$$

$$\text{Then } \int_{B_1} f = \int_A f^+ = 0 \Rightarrow f^+ = 0 \text{ a.e.}$$

$$\therefore \int_{B_2} f = \int_A f^- = 0 \Rightarrow f^- = 0 \text{ a.e.}$$

$$\Rightarrow f = 0 \text{ a.e.}$$

Prop: Let $f \in L^1(\mathbb{R}^d)$ and $\int_B f dx = 0$ for any box B , then $f = 0$ a.e.

Proof: We prove by contradiction.

WLOG, we assume $m(\{x: f(x) > 0\}) \geq a_0 > 0$.

Note: $\{x: f(x) > 0\} = \bigcup_{n=1}^{\infty} \{x: f(x) > \frac{1}{n}\}$.

$$\{x: f(x) > \frac{1}{n}\} \subset \{x: f(x) > \frac{1}{n+1}\} \Rightarrow \lim_{n \rightarrow \infty} m(\{x: f(x) > \frac{1}{n}\}) = m(\{x: f(x) > 0\}) \geq a_0.$$

Thus, $\exists n_0$ s.t. $m(\{x: f(x) > \frac{1}{n_0}\}) > \frac{1}{2}a_0$.

Take a closed set $F \subset \{x: f(x) > \frac{1}{n_0}\}$. s.t. $m(F) \geq \frac{a_0}{4}$.

$$\text{Then } \int_F f(x) dx \geq \int_F \frac{1}{n_0} dx \geq \frac{a_0}{4n_0}$$

By condition, $\int_{\mathbb{R}^d} f(x) dx = 0$. (countable additivity).

Consider $U = \mathbb{R}^d \setminus F$ It's open.

By the structure theorem of open sets, $U = \bigcup_{n=1}^{\infty} B_n$.

where B_n are almost disjoint boxes

Let $U_0 = \bigcup_{n=1}^{\infty} B_n^o$ Then $m(U \setminus U_0) = m\left(\bigcup_{n=1}^{\infty} (B_n \setminus B_n^o)\right) = 0$.

$$\begin{aligned} \text{Thus, } 0 &= \int_{\mathbb{R}^d} f(x) dx = \int_F f(x) dx + \int_{U_0} f(x) dx + \int_{U \setminus U_0} f(x) dx \\ &= \int_F f(x) dx + \sum_{i=1}^{\infty} \int_{B_i^o} f(x) dx + \int_{U \setminus U_0} f(x) dx > 0 \text{ contradiction.} \end{aligned}$$

Cor There is no measurable set $A \subset [0, 1]$

$$\text{s.t. } m(A \cap [a, b]) = \frac{b-a}{2} \quad \forall 0 \leq a \leq b \leq 1.$$

Proof: Suppose such A exists. Then for $\forall [c, d] \subset \mathbb{R}$.

$$\int_{[c, d]} (\chi_A - \frac{1}{2}\chi_{[0, 1]}) = 0.$$

$$\Rightarrow \chi_A = \frac{1}{2}\chi_{[0, 1]} \text{ a.e. contradiction.}$$

Thm Suppose $f \in L^1(A)$. Then $\forall \varepsilon > 0$. $\exists \delta > 0$ s.t.

$$\forall B \subset A \text{ measurable, } m(B) < \delta \Rightarrow \int_B |f| < \varepsilon.$$

(Absolute continuity of Lebesgue integral.)

Proof: We firstly assume $f \geq 0$.
 By non-negative theory \exists simple function φ
 $0 \leq \varphi \leq f$ s.t. $\begin{cases} \int_A (f - \varphi) dx \leq \frac{\varepsilon}{2} \\ \int_A f dx < +\infty \end{cases}$

Since φ is simple, non-negative and integrable.

One can find M s.t. $\varphi \leq M$ a.e.

$$\text{Take } g = \frac{\varepsilon}{2M}.$$

$$\int_B f dx = \int_B (f - \varphi) dx + \int_B \varphi dx < \frac{\varepsilon}{2} + M \cdot \frac{\varepsilon}{2M} = \varepsilon.$$

- For general f , we apply the above results to f^+ , f^-
 i.e. $\exists f_1$ s.t. $m(B) < \delta_1 \Rightarrow \int_B f^+ < \varepsilon/2$
 $\exists f_2$ s.t. $m(B) < \delta_2 \Rightarrow \int_B f^- < \varepsilon/2$.

Take $g = \min(f_1, f_2)$.

$$\text{Then if } m(B) < \delta. \quad \int_B |f| dx = \int_B f^+ dx + \int_B f^- dx < \varepsilon.$$

3. Convergence

$$f_n \rightarrow f \Rightarrow \int f_n \rightarrow \int f$$

Recall (Example) $f_n(x) = \begin{cases} n & 0 < x < \frac{1}{n} \\ 0 & x \leq 0 \text{ or } x \geq \frac{1}{n} \end{cases} \rightarrow f(x) = 0$.

$$\int_E f_n dx = 1 \quad \xrightarrow{} \quad \int_E f dx = 0.$$

Lebesgue Dominated Convergence Theorem (LDCT)

Suppose f_n are measurable $f_n \rightarrow f$ a.e. ($f_n \rightarrow f$ in measure)
 $\exists g \in L^1(A)$ s.t. $|f_n| \leq g$ a.e. on A .

Then $f \in L^1(A)$ and

$$\lim_{n \rightarrow \infty} \int_A f_n(x) dx = \int_A f(x) dx.$$

(In fact. $\lim_{n \rightarrow \infty} \int_A |f_n - f| dx = 0$ (i.e. $\|f_n - f\|_{L^1(\mathbb{R}^d)} \rightarrow 0$)

Lec 12

Proof. Let $A_0 = \{x : f_n(x) \rightarrow f(x)\}$.

$$A_n = \{x : f_n(x) > g(x)\}.$$

Then $m(A_0) = m(A_n) = 0$. $\Rightarrow A = \bigcup_{n=0}^{\infty} A_n$, $m(A) = 0$.

On A^c , we have $|f(x)| \leq g(x)$, $\forall x \in A^c$.

$$\Rightarrow \int_{\mathbb{R}^d} |f| dx = \int_{A^c} |f| dx \leq \int_{A^c} g dx \stackrel{(1)}{\leq} \int_{\mathbb{R}^d} g dx < +\infty.$$

$$\Rightarrow f \in L^1(\mathbb{R}^d)$$

- Apply Fatou Lemma to $f_n + g$. \leftarrow non-negative on A^c . we get

$$\begin{aligned} \int_{\mathbb{R}^d} f dx + \int_{\mathbb{R}^d} g dx &= \int_{A^c} f + g dx = \int_{A^c} \liminf_{n \rightarrow \infty} (f_n + g) dx \leq \liminf_{n \rightarrow \infty} \int_{A^c} (f_n + g) dx \\ &= \liminf_{n \rightarrow \infty} \int_{\mathbb{R}^d} f_n dx + \int_{\mathbb{R}^d} g dx \end{aligned}$$

$$\Rightarrow \int_{\mathbb{R}^d} f dx \leq \liminf_{n \rightarrow \infty} \int_{\mathbb{R}^d} f_n dx.$$

- Apply Fatou to $\{g - f_n\}$ on A^c .

$$\begin{aligned} \int_{\mathbb{R}^d} g dx - \int_{\mathbb{R}^d} f dx &= \int_{A^c} \liminf_{n \rightarrow \infty} (g - f_n) dx \leq \liminf_{n \rightarrow \infty} \int_{\mathbb{R}^d} (g - f_n) dx \\ &\leq \int_{\mathbb{R}^d} g dx - \limsup_{n \rightarrow \infty} \int_{\mathbb{R}^d} f_n dx \end{aligned}$$

$$\Rightarrow \int_{\mathbb{R}^d} f dx \geq \limsup_{n \rightarrow \infty} \int_{\mathbb{R}^d} f_n dx.$$

$$\Rightarrow \lim_{n \rightarrow \infty} \int_{\mathbb{R}^d} f_n dx = \int_{\mathbb{R}^d} f dx.$$

Cor // Under the same assumption, we have a stronger conclusion

$$\lim_{n \rightarrow \infty} \int_{\mathbb{R}^d} |f_n(x) - f(x)| dx = 0.$$

Proof. Since $|f_n - f| \rightarrow 0$ a.e. and $|f_n - f| \leq 2g$ a.e.

Apply LDCT to $\{|f_n - f|\}$, we get

$$\int_{\mathbb{R}^d} |f_n - f| dx \rightarrow \int_{\mathbb{R}^d} 0 dx = 0.$$

\ Recal

$$\|(f_n - f)\|_{L^1(\mathbb{R}^d)}$$

$(f_n, f \in L^1(\mathbb{R}^d))$

Def // We say $f_n \rightarrow f$ in L^1 norm (or $f_n \rightarrow f$ in the mean)

$$\text{if } \lim_{n \rightarrow \infty} \int_{\mathbb{R}^d} |f_n - f| dx = 0. (\|f_n - f\|_{L^1(\mathbb{R}^d)})$$

Prop // If $f_n, f \in L^1(\mathbb{R}^d)$ $f_n \rightarrow f$ in the norm.

Then $f_n \rightarrow f$ in measure $\Leftrightarrow \forall \varepsilon > 0. \lim_{n \rightarrow \infty} m(\{x : |f_n - f| > \varepsilon\}) = 0$.

Proof: Use Markov inequality $\forall \lambda > 0$

$$m(\{x : |f_n - f| > \lambda\}) \leq \frac{1}{\lambda} \int_{\mathbb{R}^d} |f_n - f| dx \rightarrow 0.$$

Therefore: in measure + dominance $\overset{f_n \rightarrow f}{\Rightarrow}$ in norm $L^1(\mathbb{R}^d)$.

Cor // If $f_n \rightarrow f$ in norm. \exists sequence $\xrightarrow{f_n} f$ a.e.

Rmk: uniform convergence $\not\Rightarrow$ convergence in L^1 -norm.

(None of convergence modes in Lee 7 implies convergence in L^1 -norm)

e.g. $f_n = \frac{1}{n} \chi_{[0, n]}$ $\rightarrow f = 0$ uniformly. $\int_{\mathbb{R}^d} |f_n - f| dx = 1 \not\rightarrow 0$

Thm // (Completeness of $L^1(\mathbb{R}^d)$) (Riesz-Fischer)

"Cauchy sequence converges"

$\{x_n\}. \forall \varepsilon > 0. \exists N$ s.t. $\forall n, m > N. |x_m - x_n| < \varepsilon$.

Def: A sequence $f_n \in L^1(\mathbb{R}^d)$ is a cauchy sequence in $L^1(\mathbb{R}^d)$

if $\forall \varepsilon > 0. \exists N$ s.t. $\forall n, m > N$.

$$\|f_m - f_n\|_{L^1} = \int_{\mathbb{R}^d} |f_m(x) - f_n(x)| dx < \varepsilon.$$

↑ norm → distance

For any Cauchy sequence $f_n \in L^1(\mathbb{R}^d)$, $\exists f \in L^1(\mathbb{R}^d)$ s.t:

$f_n \rightarrow f$ in L^1 -norm.

Proof: First we take a subsequence f_{n_k} s.t:

$$\|f_{n_{k+1}} - f_{n_k}\|_{L^1(\mathbb{R}^d)} < \frac{1}{2^k}$$

Let $g_N(x) = \sum_{k=1}^N |f_{n_{k+1}}(x) - f_{n_k}(x)| + |f_{n_1}(x)|$ Then

g_1, g_2, \dots are non-negative, increasing

$$\Rightarrow g_N \rightarrow g \in L^1(R^d)$$

$$[\int_{R^d} g dx = \lim_{N \rightarrow \infty} \int_{R^d} g_N dx \leq \int_{R^d} |f_{n_1}| dx + \sum_{k=1}^{\infty} \frac{1}{2^k} < \infty]$$

Let $h_N(x) = f_{n_1}(x) + \sum_{k=1}^N (|f_{n_{k+1}}(x) - f_{n_k}(x)|) (= f_{n_{N+1}}(x))$

Then $|h_N(x)| \leq g(x)$

$\Rightarrow \forall x$. Series $f_{n_1}(x) + \sum_{k=1}^{\infty} (f_{n_{k+1}}(x) - f_{n_k}(x))$ converges to some $f(x)$.

$f(x) = \lim_{N \rightarrow \infty} (f_{n_1}(x) + \sum_{k=1}^N (f_{n_{k+1}}(x) - f_{n_k}(x)))$ is measurable
 $= \lim_{N \rightarrow \infty} h_N(x).$

• Apply LDCT. $f \in L^1(R^d), \lim_{N \rightarrow \infty} \int_{R^d} |f - h_N| dx = 0.$
 $\Rightarrow \lim_{N \rightarrow \infty} \int_{R^d} |f - f_{n_{N+1}}| dx = 0.$

i.e. $\forall \epsilon > 0 \exists K$ s.t. $\forall N > K$.

$$\|f_{n_{N+1}} - f\|_{L^1(R^d)} < \frac{\epsilon}{2}.$$

By def of Cauchy sequence $\exists M$ s.t. $\forall n, m > M$.

$$\|f_n - f_m\|_{L^1(R^d)} < \frac{\epsilon}{2}$$

Take $L = \max\{K, M\}$, then $\forall n > L$.

$$\|f_n - f\|_{L^1(R^d)} \leq \|f_n - f_{n_{L+1}}\|_{L^1(R^d)} + \|f_{n_{L+1}} - f\|_{L^1(R^d)} < \epsilon.$$

Proof of LDCT (in measure)

1st: By Riesz, \exists subseq. $f_{n_k} \rightarrow f$ a.e.

• By LDCT $f \in L^1(R^d)$ and $\|f_{n_k} - f\|_{L^1} \rightarrow 0$ (converges in L^1 norm)

• Suppose $\|f_n - f\|_{L^1} \not\rightarrow 0$. Then $\exists \epsilon_0$ s.t. $\|f_{n_k} - f\|_{L^1} \geq \epsilon_0 \quad \forall k$.

Note: $f_{n_k} \rightarrow f$ in measure $\Rightarrow \exists f'_{n_k}$ s.t. $\|f_{n_k} - f'\|_{L^1} \rightarrow 0$ subseq

Contradiction

2nd proof: ① $f_n - f \in L^1(\mathbb{R}^d)$.

② WANT $\int_{\mathbb{R}^d} |f_n - f| dx \rightarrow 0$. ($< \varepsilon$, $\forall \varepsilon > 0$).

$$\begin{aligned}
 &= \int_{\mathbb{R}^d} |f_n - f| dx \stackrel{\leq \varepsilon}{\approx} + \int_{\{|f_n - f| < \varepsilon\}} |f_n - f| dx \stackrel{\leq \varepsilon}{\approx} \\
 &\quad \xrightarrow{\substack{\{f_n - f < \varepsilon\} \cap B(0, n) \\ + \int_{\{f_n - f < \varepsilon\} \cap B(0, n)^c} |f_n - f| dx}} \\
 &\quad \int_{B(0, n)} |f| dx \rightarrow \underbrace{\int_{\mathbb{R}^d} |h| dx}_{< +\infty} \stackrel{\leq \varepsilon}{\approx}
 \end{aligned}$$

Lec 13

1 Countable additivity (w.r.t Domain and Function).

Prop: || Suppose $f \in L^1(A)$, $A = \bigcup_{i=1}^{\infty} A_i$. $A_i \cap A_j \neq \emptyset$ Then $\int_A f dx = \sum_{i=1}^{\infty} \int_{A_i} f dx$

2nd proof: Let $\tilde{f}_i = f|_{A_i}$. $f_n = \sum_{i=1}^n \tilde{f}_i$ Then $f_n \rightarrow f$.
 $|f_n| \leq |f| \in L^1(A)$

By LDCT

$$\int_A f dx = \lim_{n \rightarrow \infty} \int_A f_n dx = \lim_{n \rightarrow \infty} \sum_{i=1}^n \int_{A_i} f_n dx = \sum_{i=1}^{\infty} \int_{A_i} f dx.$$

$$\sum_{n=1}^{\infty} \int_A |f_n| dx < \infty.$$

Prop: Let $f_n \in L^1(A)$ Then $f \in L^1(A)$ s.t. $\sum_{n=1}^{\infty} f_n$ converges to f a.e.
and $\int_A f dx = \sum_{n=1}^{\infty} \int_A f_n dx$.

Proof: By non-negative theory $\sum_{n=1}^{\infty} \int_A |f_n| dx = \int_A \sum_{n=1}^{\infty} |f_n| dx < +\infty$.
 $\Rightarrow F = \sum_{n=1}^{\infty} |f_n| \in L^1$.

$\Rightarrow F$ is a.e. finite

$\Rightarrow \sum_{n=1}^{\infty} f_n$ a.e. absolutely convergent

$\Rightarrow \sum_{n=1}^{\infty} f_n$ a.e. converges to a function f .

Note: $f \stackrel{\text{a.e.}}{=} \lim_{N \rightarrow \infty} \sum_{n=1}^N f_n \rightarrow f$ is measurable.

Since $|f| \leq F \Rightarrow f \in L^1(A)$.

Let $g_N = \sum_{n=1}^N f_n$. Then $g_N \rightarrow f$. a.e. and $|g_N| \leq F$.

By LDCT.

$$\int_A f dx = \lim_{N \rightarrow \infty} \int_A g_N dx = \lim_{N \rightarrow \infty} \sum_{n=1}^N \int_A f_n dx = \sum_{n=1}^{\infty} \int_A f_n dx.$$

2. Riemann Integral v.s. Lebesgue integral

- Let $f: [a,b] \rightarrow \mathbb{R}$ be a bounded function

Recall for any partition P of $[a,b]$

$$S_P(f) = \sum_{k=1}^n \sup_{x_{k-1} \leq x \leq x_k} f(x) \cdot (x_k - x_{k-1}) \quad \text{Darboux upper sum}$$

$$s_P(f) = \sum_{k=1}^n \inf_{x_{k-1} \leq x \leq x_k} f(x) \cdot (x_k - x_{k-1}) \quad \text{Darboux lower sum}$$

$$\begin{aligned} \overline{\int_a^b f dx} &= \inf_P \bar{S}_P(f) \quad \text{--- Darboux upper integral} \\ \underline{\int_a^b f dx} &= \sup_P S_P(f) \quad \text{--- Darboux lower integral} \end{aligned}$$

f is Riemann integral $\Leftrightarrow \overline{\int_a^b f dx} = \underline{\int_a^b f dx} =: \int_a^b f dx.$

Now given a partition P , we can define step function

$$\bar{\Phi}_P(f) = \sum_{k=1}^n \sup_{[x_{k-1}, x_k]} f \cdot \chi_{[x_{k-1}, x_k]} \quad (= \bar{S}_P(f))$$

$$\underline{\Phi}_P(f) = \sum_{k=1}^n \inf_{[x_{k-1}, x_k]} f \cdot \chi_{[x_{k-1}, x_k]} \quad (= S_P(f))$$

By the trick of "common refinement", we can find P_n with

$P_1 \leq P_2 \leq \dots$ (P_{n+1} is a refinement of P_n)

$$\text{s.t.: } \int_a^b f dx = \lim_{n \rightarrow \infty} \bar{S}_{P_n}(f) = \lim_{n \rightarrow \infty} \int_{[a,b]} \bar{\Phi}_{P_n}(f) dx.$$

$$\int_a^b f dx = \lim_{n \rightarrow \infty} \bar{S}_{P_n}(f) = \lim_{n \rightarrow \infty} \int_{[a,b]} \bar{\Phi}_{P_n}(f) dx.$$

One can check

$$\begin{aligned} \bar{\Phi}_{P_n}(f) &\nearrow \bar{\Phi}(f) \\ \underline{\Phi}_{P_n}(f) &\nearrow \underline{\Phi}(f) \end{aligned} \quad \left. \begin{array}{l} \text{measurable} \end{array} \right\} \underline{\Phi}(f) \leq f \leq \bar{\Phi}(f)$$

Now we have

$$\begin{aligned} f \text{ is Riemann integral} &\Leftrightarrow \lim_{n \rightarrow \infty} \int_{[a,b]} (\bar{\Phi}_{P_n}(f) - \underline{\Phi}_{P_n}(f)) dx = 0. \\ &\Leftrightarrow \int_{[a,b]} (\bar{\Phi}(f) - \underline{\Phi}(f)) dx = 0. \\ &\Leftrightarrow \underline{\Phi}(f) = \bar{\Phi}(f) \text{ a.e. in } [a,b]. \\ &\quad (\Leftrightarrow f \text{ is discontinuous only on a set measure 0}) \end{aligned}$$

Prop. || If a bounded function $f : [a,b] \rightarrow \mathbb{R}$ is Riemann integrable then $f \in C^1([a,b])$ and $\int_{[a,b]} f dx = \int_a^b f(x) dx$.

Proof: By LDCIT. $\underline{\Phi}(f) \in C^1([a,b]) \Rightarrow f \in C^1([a,b])$

$$\text{and } \int_{[a,b]} f dx = \int_{[a,b]} \underline{\Phi}(f) dx = \lim_{n \rightarrow \infty} \int_{[a,b]} \underline{\Phi}_{P_n}(f) dx.$$

$$= \int_a^b f(x) dx = \int_a^b f(x) dx.$$

Rmk There exists functions f so that the improper integral $\int_a^b f(x) dx$ (in Riemann sense) exists, while f is not Lebesgue integrable.
 $f: [0, +\infty) \rightarrow \mathbb{R}$ $f(x) = \sum_{n=1}^{\infty} \frac{(-1)^n}{n} \chi_{(n-1, n)}$. $\Rightarrow \int_0^{+\infty} f(x) dx = \sum_{n=1}^{\infty} \frac{(-1)^n}{n} = -\ln 2$.
but $\int_{[0, +\infty)} |f| dx = +\infty$.

3. Integrals with parameters:

$$\bar{F}(t) = \int_A f(t, x) dx.$$

Prop. Let $T \subset \mathbb{R}^{d_1}$, $A \subset \mathbb{R}^{d_2}$ be measurable sets, $f: T \times A \rightarrow \mathbb{R}$ be a function.
Suppose ① $\forall t \in T$ $f_t(x) = f(t, x)$ is a measurable function on A .
② $\forall x \in A$ $f^{(x)}(t) = f(t, x)$ is a continuous function on T .
Moreover, suppose $\exists g \in L'(A)$ s.t. $|f_t(x)| \leq g(x)$ a.e. $x \in A$ $\forall t \in T$.
Then for $\forall t \in T$, $f_t \in L'(A)$ and $\bar{F}(t) := \int_A f(t, x) dx$ is continuous w.r.t. $t \in T$.
Proof: Suppose $t \in T$ and $t_n \rightarrow t \in T$. Then
 $\forall x \in A$, $f_{t_n}(x) = f_{t_n}(x) \rightarrow f_{t_0}(x) = f(t_0, x)$ AND $|f_{t_n}(x)| \leq g(x) \in L'$.
By LDCT, we have $\bar{F}(t_n) \rightarrow \bar{F}(t_0)$.

Prop. Let $f: (a, b) \times A \rightarrow \mathbb{R}$ be a function s.t.

① $\forall t \in (a, b)$, $f_t(x) \in L'(A)$

② $\forall x \in A$: $f^{(x)}(t)$ is differentiable.

Suppose $\exists g \in L'(A)$ s.t. $|\frac{d}{dt} f(t, x)| \leq g(x)$ ($\forall t \in (a, b)$) $\forall x \in A$.

Then $\forall t \in (a, b)$

$$\frac{d}{dt} \int_A f(t, x) dx = \int_A \frac{d}{dt} f(t, x) dx.$$

Proof: For any $(t, x) \in (a, b) \times A$, one has

$$\frac{d}{dt} f(t, x) = \lim_{h \rightarrow 0} \frac{f(t+h, x) - f(t, x)}{h}.$$

e.o)

By mean value theorem,

$$\left| \frac{f(t+h, x) - f(t, x)}{h} \right| \leq g(x). \quad \forall x \in A. \quad h \rightarrow 0$$

by LDCT

$$\begin{aligned} \frac{d}{dt} \int_A f(t, x) dx &= \lim_{h \rightarrow 0} \int_A \frac{f(t+h, x) - f(t, x)}{h} dx \\ &= \int_A \frac{d}{dt} f(t, x) dx. \end{aligned}$$

4. The theorem of Fubini
 Suppose $f = f(x, y)$ be a "2-variable" function on $A \times B \subset R^{d_1} \times R^{d_2}$.
 Question: $\int_{A \times B} f(x, y) dx dy$ $\neq \int_A \left(\int_B f(x, y) dx \right) dy$.
 Example: $f(x, y) = \frac{x^2 - y^2}{(x+y)^2}$ $\int_{(x,y) \in [0,1] \times [0,1]} \int_A \left(\int_B f(x, y) dx \right) dy$.
 By using formulas $\int_{x^2 + a^2} \frac{dx}{x^2 + a^2} = \frac{1}{a} \arctan \frac{x}{a} + C$, $\int_{(x+y)^2} \frac{dx}{(x+y)^2} = \frac{x}{2a(x+a^2)} + \frac{1}{2a} \int \frac{dx}{x^2 + a^2}$
 $\int_0^1 \left(\int_0^1 f(x, y) dx \right) dy = \int_0^1 \left(-\frac{1}{1+y^2} \right) dy = -\frac{\pi}{4}$.
 One can show $\int_0^1 \int_0^1 f(x, y) dy dx = \int_0^1 \frac{1}{1+x^2} dx = \frac{\pi}{4}$.
 $\int_0^1 \left(\int_0^1 |f(x, y)| dx \right) dy = +\infty \rightarrow$ Not integral.

[Reason]: $\int_0^1 \left(\int_0^1 |f(x, y)| dx \right) dy \in L^1(A \times B)$

Thm (Fubini) Let $A \subset R^{d_1}, B \subset R^{d_2}$ be measurable. $f = f(x, y) \in L^1(A \times B)$
 Then (i) For a.e. $y \in B$. $f_y(x) := f(x, y)$, then $f_y \in L^1(A)$
 (ii) As a function of y $F(y) := \int_A f(x, y) dx$, then $F \in L^1(B)$
 (iii) $\int_{A \times B} f(x, y) dx dy = \int_B \left(\int_A f(x, y) dx \right) dy + \int_A \left(\int_B f(x, y) dy \right) dx$.

Rmk: By zero extension, we may assume $A = R^{d_1}, B = R^{d_2}$
 For (3) it's enough to prove the first = "

Special \rightarrow General

Proof: WLOG, we assume $A = R^{d_1}, B = R^{d_2}$.
 For simplicity. We let $\mathcal{F} = \{f \in L^1(R^{d_1} \times R^{d_2}) : f \text{ satisfies } (i), (ii), (iii)\}$.
 (Path of proof: $X_{\text{box}} \rightsquigarrow X_{\text{open}} \rightsquigarrow X_{\text{measurable}}$)

[Step 1] Let $f = \chi_{B_1 \times B_2}$, where $B_1 \subset R^{d_1}, B_2 \subset R^{d_2}$ are boxes.
 Then $f \in \mathcal{F}$.

To see this: (1) $f_y(x) = \chi_{B_2}(y) \cdot \chi_{B_1}(x) \in L^1(R^{d_1})$

(2) $F(y) = \int_{R^{d_1}} f(x, y) dx = \int_{R^{d_1}} \chi_{B_2}(y) \chi_{B_1}(x) dx = \chi_{B_2}(y) \cdot m(B_1) \in L^1(R^{d_2})$

(3) $\int_{R^{d_1} \times R^{d_2}} f(x, y) dx dy = m(B_1 \times B_2) = m(B_1) \cdot m(B_2)$
 $= \int_{R^{d_2}} \left(\int_{R^{d_1}} f(x, y) dx \right) dy = \int_{R^{d_2}} m(B_1) \chi_{B_2}(y) dy = m(B_1) \cdot m(B_2)$

4. The theorems of Fubini

Suppose $f = f(x, y)$ be a "2-variable" function on $A \times B \subset R^{d_1} \times R^{d_2}$.

Question $\int_{A \times B} f(x, y) dx dy \Rightarrow \int_B (\int_A f(x, y) dx) dy$.

Example: $f(x, y) = \frac{x^2 - y^2}{(x^2 + y^2)^2}$ $(x, y) \in [0, 1] \times [0, 1]$.

By using formulas $\int \frac{dx}{x^2 + a^2} = \frac{1}{a} \arctan \frac{x}{a} + C$, $\int \frac{dx}{(x+a)^2} = \frac{x}{2a(x+a^2)} + \frac{1}{2a} \int \frac{dx}{x^2 + a^2}$

One can show $\int_0^1 (\int_0^1 f(x, y) dx) dy = \int_0^1 (-\frac{1}{1+y^2}) dy = -\frac{\pi}{4}$

$\int_0^1 (\int_0^1 f(x, y) dy) dx = \int_0^1 \frac{1}{1+x^2} dx = \frac{\pi}{4}$

[Reason]: $\int_1^1 (\int_0^1 |f(x, y)| dx) dy = +\infty \rightarrow$ not integral.

Thm (Fubini) Let $A \subset R^{d_1}$, $B \subset R^{d_2}$ be measurable. $f = f(x, y) \in L^1(A \times B)$

Then (i) For a.e. $y \in B$. $f_y(x) := f(x, y)$, then $f_y \in L^1(A)$

(ii) As a function of y $F(y) := \int_A f(x, y) dx$, then $F \in L^1(B)$

(iii) $\int_{A \times B} f(x, y) dx dy = \int_B (\int_A f(x, y) dx) dy = \int_A (\int_B f(x, y) dy) dx$

Rmk.: By zero extension, we may assume $A = R^{d_1}$, $B = R^{d_2}$

* For (iii) it's enough to prove the first " = "

Special \rightarrow General

Proof: WLOG, we assume $A = R^{d_1}$, $B = R^{d_2}$.

For simplicity. We let $\Omega_f = \{f \in L^1(R^{d_1} \times R^{d_2}) : f \text{ satisfies } X_G, (i), (ii), (iii)\}$.
 (Path of proof: $X_{\text{Box}} \rightsquigarrow X_{\text{open}} \rightsquigarrow X_{\text{measur.}} \rightsquigarrow X_{\text{measurable}}$.
 $\rightarrow f \in \text{simple.} \rightarrow f \in \Omega_f$)

[Step 1] Let $f = \chi_{B_1 \times B_2}$, where $B_1 \subset R^{d_1}$, $B_2 \subset R^{d_2}$ are boxes
 Then $f \in \Omega_f$.

To see this: (i) $f_y(x) = \chi_{B_2}(y) \cdot \chi_{B_1}(x) \in L^1(R^{d_1})$

(ii) $F(y) = \int_{R^{d_1}} f(x, y) dx = \int_{R^{d_1}} \chi_{B_2}(y) \chi_{B_1}(x) dx = \chi_{B_2}(y) \cdot m(B_1) \in L^1(R^{d_2})$

(iii) $\int_{R^{d_1} \times R^{d_2}} f(x, y) dx dy = m(B_1 \times B_2) = m(B_1) \times m(B_2)$
 $\qquad \qquad \qquad = \int_{R^{d_2}} (\int_{R^{d_1}} f(x, y) dx) dy = \int_{R^{d_2}} m(B_1) \chi_{B_2}(y) dy = m(B_1) \times m(B_2)$

Step 2 If $f_1, f_2 \in \mathcal{F}$, then $c_1 f_1 + c_2 f_2 \in \mathcal{F}$

(If $f_1, \dots, f_n \in \mathcal{F}$, $\sum_{i=1}^n c_i f_i \in \mathcal{F}$) finite case.

This follows from the linearity of Lebesgue integral.

Step 3 Suppose either if $f_n \nearrow f$ $f_n \in \mathcal{F}$ or if $f_n \searrow f$ $f_n \in \mathcal{F}$.

Proof. enough to prove (a)

Consider $f_n - f_1 \nearrow f - f_1$ each $f_n - f_1 \geq 0$.

By monotone convergence thm $f \in L^1(R^{d_1} \times R^{d_2})$

$$\int_{R^{d_1} \times R^{d_2}} f_n(x, y) dx dy \rightarrow \int_{R^{d_1} \times R^{d_2}} f(x, y) dx dy$$

(1a) For a.e. $y \in R^{d_2}$ we have $(f_n)_y \in L^1(R^{d_1})$ $(f_n)_y \nearrow f_y$.

Moreover $f_y \geq (f_1)_y$. By monotone convergence.

$$F_n(y) = \int_{R^{d_1}} f_n(x, y) dx \xrightarrow{(m.c.)} \int_{R^{d_1}} f_y(x) dx := F(y) \quad \text{still need to prove } f_y \in L^1(R^{d_1})$$

(2) We have $F_n \nearrow F$, $F_n \in L^1(R^{d_2})$

$$\int_{R^{d_2}} F_n(y) dy \rightarrow \int_{R^{d_2}} F(y) dy$$

It follows that $\int_{R^{d_2}} F(y) dy = \int_{R^{d_1} \times R^{d_2}} f(x, y) dx dy < \infty$.

$$\text{So } F \in L^1(R^{d_2})$$

(1b) So for a.e. y , $F(y)$ is finite \Rightarrow for a.e. y , $f_y \in L^1(R^{d_1})$.

$$(3) \int_{R^{d_1} \times R^{d_2}} f(x, y) dx dy = \int_{R^{d_2}} F(y) dy = \int_{R^{d_2}} \left(\int_{R^{d_1}} f(x, y) dx \right) dy.$$

Step 4 If E is a subset of the boundary of a $\boxed{\text{box}}$ ($\Rightarrow \underline{\text{meas}}(E) = 0$) bounded

then $\chi_E \in \mathcal{F}$

Step 5 If V is a bounded open set and $f = \chi_u$, then $f \in \mathcal{F}$.

We can write $V = \bigcup_{k=1}^{\infty} \overline{B_k} = \bigcup_{k=1}^{\infty} B_k$, where $\overline{B_k}$'s are disjoint boxes, and B_k 's are disjoint s.t. $\overline{B_k} \subset B_k \subset B_k$

By step 1, 2, 4. $\chi_{B_k} \in \mathcal{F}$.

By step 2 $\chi_{\bigcup_{i=1}^k B_i} \in \mathcal{F}$.

By step 3. $\chi_u \in \mathcal{F}$.

Step 7

For any measure σ set $E \in \mathcal{F}$.

Since $E = \bigcup_{n=1}^{\infty} (B_{1,0,n} \cap E)$ while each $B_{1,0,n} \cap E$ is bounded

and of measure 0, So. one can apply step 3

Step 8

If $E \in \mathcal{L}$, $m(E) < \infty$. then $\chi_E \in \mathcal{F}$.

In fact, $E \in \mathcal{L}, m(E) < \infty \Rightarrow E = G \setminus N$. Where $G = \bigcap_{n=1}^{\infty} U_n$ is
a G_δ -set. each $m(U_n) < \infty$. and $m(N) = 0$.

Apply Step 5, 3, 7, 2. $\chi_E \in \mathcal{F}$.

Step 9

Any integral simple function $\in \mathcal{F}$.

Apply step 2, 8.

Step 10

Any integrable function $f \in \mathcal{F}$

$f = f^+ - f^-$. $f^+, f^- \geq 0$. and integral

By Step 3, 9. $f^+, f^- \in \mathcal{F}$.

By step 2 $f \in \mathcal{F}$.

Lee 14.

✓ Tonelli: Suppose f is measurable and ~~non-negative~~ on $A \times B$.

Then
symmetry
1) for a.e. y , $f(x, y)$ is measurable on A
2) The function $F(y) = \int_A f(x, y) dx$ is measurable on B
3) $\int_{A \times B} f(x, y) dx dy = \int_B (\int_A f(x, y) dx) dy (= \int_A (\int_B f(x, y) dy) dx)$
[may be $+\infty$]

Idea: Try to find a sequence of $f_n \in L^1(A \times B) \rightarrow f$ (to use MCT)
How? Cut both the domain and range.

Proof: For any n , we define $f_n(x) = \begin{cases} f(x) & \text{if } (x, y) \in B(0, n), f(x) > n \\ n & \text{if } (x, y) \in B(0, n), f(x) \leq n \\ 0 & \text{if } (x, y) \notin B(0, n) \end{cases}$
Then $0 \leq f_1 \leq f_2 \leq \dots \nearrow f$
 $f_n \in L^1(A \times B)$

By Fubini (1). $\exists B_n$ with $m(B_n) = 0$. s.t. $f_n(x, y) \in L^1(A)$ for $y \in B \setminus B_n$

$\rightarrow f_n(x, y)$ is measurable

\Rightarrow for $y \in B \setminus (\bigcup_{i=1}^n B_i)$ $f_n(x, y) \rightarrow f(x, y) \Rightarrow f(x, y) \stackrel{(1)}{\text{measurable}}$
Since $f_n(\cdot, y) \rightarrow f(\cdot, y)$ By MCT.

$$F_n(\cdot, y) = \int_A f_n(x, y) dx \nearrow \int_A f(x, y) dx = F(y) \quad (2)$$

By Fubini (2) each F_n is measurable $\Rightarrow F$ is measurable.

By MCT for F , we get $\int_B (\int_A f_n(x, y) dx) dy \rightarrow \int_B (\int_A f(x, y) dx) dy$

By MCT for f_n w.r.t. (x, y) $\int_{A \times B} f_n(x, y) dx dy \rightarrow \int_{A \times B} f(x, y) dx dy$ \checkmark (3)

Rank: In applications, one may need to combine the Tonelli and Fubini:

i.e. First apply Tonelli to $|f| \stackrel{>0}{\rightarrow}$ compute $\int_{A \times B} |f| = \int_B (\int_A |f| dx) dy \neq$

If yes, the $f \in L^1(A \times B)$. So can apply Fubini to compute $\int_{A \times B} f d$

Corollary If $A \in \mathcal{R}^{d_1}$, $B \in \mathcal{R}^{d_2}$ are measurable, then $A \times B$ is measurable
and $m(A \times B) = m(A) \times m(B)$ The proof is pure measure theoretic

Proof: Assuming first half, then apply Tonelli to $\chi_{A \times B}$
to get $m(A \times B) = \int_{\mathcal{R}^{d_1} \times \mathcal{R}^{d_2}} \chi_{A \times B} dx dy$
 $= \int_{\mathcal{R}^{d_2}} m(A) \chi_B(y) dy = m(A)m(B).$

To prove the first half: $A = \bigcup_{n=0}^{\infty} A_n$ where $m(A_0) = 0$ and each A_n is closed.

Want: $A \times B = (\bigcup_{n=0}^{\infty} A_n) \times (\bigcup_{m=0}^{\infty} B_m) = \bigcup_{n,m=0}^{\infty} (A_n \times B_m)$ where $m(B_0) = 0$ and each B_m is closed

- ① For $m, n \geq 1$. A_n, B_m are closed $\Rightarrow A_n \times B_m$ is closed.
- ② It remains to prove $A_0 \times B_m, A_m \times B_0$ are measurable.

[Observation: They should be measurable]

Since $m(A_0) = 0$. one can find $A_0 \subset \bigcup_{i=1}^{\infty} C_i$ where C_i is a box, $\sum_{i=1}^{\infty} m(C_i) < \varepsilon$.

Similarly, $B_m \subset \bigcup_{j=1}^{\infty} D_j^{(m)}$ where $D_j^{(m)}$ are boxes s.t. $\sum_{j=1}^{\infty} m(D_j^{(m)}) < m(B_m) + \varepsilon$.

$$\begin{aligned} \sim m(A_0 \times B_m) &\leq m\left(\bigcup_{i,j} C_i \times D_j^{(m)}\right) \leq \sum_{i,j} m(C_i) \times m(D_j^{(m)}) = \left(\sum_i m(C_i)\right) \left(\sum_j m(D_j^{(m)})\right) \\ &< \varepsilon(m(B_m) + \varepsilon) \xrightarrow[\varepsilon \rightarrow 0]{} 0. \end{aligned}$$

$A_n^{\text{new}} = \overline{\left(\bigcup_{j=1}^n A_j\right) \cap B(n, n)}$ closed, bounded $\bigcup A_n^{\text{new}} = \bigcup A_n = A$.
 $B = \bigcup_{i=1}^{\infty} B_i^{\text{new}}$ the same way.

Recall $f \geq 0$ measurable. a.e. finite.

$$\int_A f(x) dx = \int_0^{\infty} m(\{x : f(x) > t\}) dt$$

Riemann

Prop: || Suppose f is measurable, then for $1 \leq p < \infty$,

$$\int_A |f(x)|^p dx = p \int_0^\infty t^{p-1} m(\{x: |f(x)| > t\}) dt.$$

Proof: Let $F(t, x) = \chi_{\{|f(x)| > t\}}$.

$$\begin{aligned}\int_A |f(x)|^p dx &= \int_A \left(\int_0^{|f(x)|} pt^{p-1} dt \right) dx \\ &= \int_A \int_0^\infty pt^{p-1} F(t, x) dt dx \\ &= \int_0^\infty \left(\int_A pt^{p-1} F(t, x) dx \right) dt \\ &= p \int_0^\infty t^{p-1} m(\{x: |f(x)| > t\}) dt.\end{aligned}$$

Def: || Given two measurable function on \mathbb{R}^d . we define their Convolution to be

$$(f * g)(x) = \int_{\mathbb{R}^d} f(x-y) g(y) dy.$$

If the integral exists.

Rank: $(g * f)(x) = \int_{\mathbb{R}^d} f(y) g(x-y) dy = \int_{\mathbb{R}^d} f(x-y) g(y) dy = (f * g)(x).$

Prop: || If $f, g \in L^1(\mathbb{R}^d)$ Then

(1) for a.e $x \in \mathbb{R}^d$ $(f * g)(x)$ exists.

(2) $f * g \in L^1(\mathbb{R}^d)$

(3) $\int_{\mathbb{R}^d} |f * g|(x) dx \leq \left(\int_{\mathbb{R}^d} |f(x)| dx \right) \cdot \left(\int_{\mathbb{R}^d} |g(y)| dy \right)$

Proof: we prove (3) (\Rightarrow (2) \Rightarrow (1)).

First assume $f, g \geq 0$. Then

$$\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(x-y) g(y) dy dx = \int_{\mathbb{R}^d} \left(\int_{\mathbb{R}^d} f(x-y) dx \right) g(y) dy$$

$$= \int_{\mathbb{R}^d} g(y) \left(\int_{\mathbb{R}^d} f(x) dx \right) dy = \int_{\mathbb{R}^d} f(x) dx \cdot \int_{\mathbb{R}^d} g(y) dy$$

For general, we only need to prove

$$|f * g| \leq |f| * |g|.$$

To more properties of convolution, we need

Thm || If $f \in L^1(\mathbb{R}^d)$, then $\int_{\mathbb{R}^d} |f(x+h) - f(x)| dx \xrightarrow{h \rightarrow 0} 0$

(integral function is almost continuous in sense of mean)

Proof: Take K s.t: $\int_{\mathbb{R}^d \setminus B(0, K)} |f(x)| dx < \varepsilon$ ($\leftarrow \lim_{K \rightarrow \infty} \int_{\mathbb{R}^d \setminus B(0, K)} |f(x)| dx = \int_{\mathbb{R}^d} |f(x)| dx$)
Take $g \in L^1(\mathbb{R}^d)$ s.t: $\|g - f\|_{L^1(\mathbb{R}^d)} < \varepsilon$
continuous

$$\begin{aligned} \Rightarrow \int_{\mathbb{R}^d} & |f(x+h) - f(x)| dx = \int_{\mathbb{R}^d} |f(x+h) - f(x)| dx \\ & + \int_{B(0, K+1)} |f(x+h) - f(x)| dx \quad \text{ensure } f(x+h) \in B(0, K) \ (h \rightarrow 0) \\ & \leq \varepsilon + \varepsilon + \|f(x+h) - g(x+h) + g(x+h) - g(x) + g(x) - f(x)\|_{L^1(B(0, K+1))} \\ & \leq 2\varepsilon + \|f(x+h) - g(x+h)\|_{L^1(\mathbb{R}^d)} + \|g(x+h) - g(x)\|_{L^1(B(0, K+1))} + \|g(x) - f(x)\|_{L^1(\mathbb{R}^d)} \\ & \leq 4\varepsilon + \|g(x+h) - g(x)\|_{L^1(B(0, K+1))} < 5\varepsilon. \end{aligned}$$

uniformly continuous in $B(0, K+1)$

Cor || $f \in L^1(\mathbb{R}^d)$. g is bounded, measurable, then

$f * g$ is uniformly continuous on \mathbb{R}^d

Proof: Let $M = \sup |g| < \infty$.

$$\begin{aligned} |(f * g)(x+h) - (f * g)(x)| &= \left| \int_{\mathbb{R}^d} f(x+h-t) g(t) dt - \int_{\mathbb{R}^d} f(x-t) g(t) dt \right| \\ &\leq \int_{\mathbb{R}^d} |f(x+h-t) - f(x-t)| \cdot |g(t)| dt \\ &\leq M \int_{\mathbb{R}^d} |f(x+h-t) - f(x-t)| dt \\ &= M \underbrace{\int_{\mathbb{R}^d} |f(t+h) - f(t)| dt}_{\xrightarrow{h \rightarrow 0}}. \end{aligned}$$

Cor :|| Suppose $f \in L^1(\mathbb{R}^d)$ $g \in C^k(\mathbb{R}^d)$ and g is compact supported
Then $f * g \in C^k(\mathbb{R}^d)$

Proof: $\frac{\partial}{\partial x_i} (f * g) = \frac{\partial}{\partial x_i} \int_{\mathbb{R}^d} f(y) g(x-y) dy$

$$\xrightarrow{\text{Prop last time}} \int_{\mathbb{R}^d} f(y) \frac{\partial}{\partial x_i} g(x-y) dy = f * \underbrace{\frac{\partial}{\partial x_i} g}_{\substack{\text{y bounded} \\ \text{continuous}}}.$$

$\exists \eta_h$ s.t. $f * \eta_h \in C^\infty$.

and $\|f * \eta_h - f\|_{L^1} \rightarrow 0$ ($h \rightarrow 0$).

$\cdot L: \mathbb{R}^d \rightarrow \mathbb{R}^d$ linear $\Rightarrow m(L(A)) = |\det L| m(A)$.

$\cdot f: \mathbb{R}^d \rightarrow \mathbb{R}$ non-negative, measurable $\Rightarrow \int_{\mathbb{R}^d} f(x) dx = m(\{(x,y) \in \mathbb{R}^{d+1} : 0 \leq y \leq f(x)\})$

Lec 15.

1 Measurable space.

i.e. for subsets of X , we can "tell how large it is" ^{mid-term}

Want: define measure (and then integral) on an abstract space
A set X .
We have seen:

If we want to keep nice property.

[e.g. countable additivity, translation-inv, $m([0,1]) = 1 \Rightarrow m \text{ on } \mathbb{Z}^d$]

Then one can't define measure on all sets of X .

→ We only pick some nice subsets of X to define measure

Recall: In Lebesgue theory, we define measure for $A \in \mathcal{L}$.
(all Lebesgue measurable subsets in \mathbb{R}^d).

(1) $\emptyset \in \mathcal{L}$.

(2) $A \in \mathcal{L}, \text{then } A^c \in \mathcal{L}^c$.

(3) $A_1, A_2 \in \mathcal{L} \Rightarrow A_1 \cap A_2, A_1 \cup A_2, A_1 \setminus A_2, A_1 \Delta A_2 \in \mathcal{L}$.

(4) $A_1, A_2, \dots \in \mathcal{L} \Rightarrow \bigcup_{i=1}^{\infty} A_i, \bigcap_{i=1}^{\infty} A_i \in \mathcal{L}$.

(5) open, closed, compact, F_σ , $G_\delta, \dots \in \mathcal{L}$. drop it here
countable \cup ← the conceptions are not defined on abstract set

Recall: $\bigcap_{i=1}^{\infty} A_i \in \mathcal{L} \Leftarrow (2), (4) \text{ 1st half } (\bigcap_{i=1}^{\infty} A_n)^c = \left(\bigcup_{i=1}^{\infty} A_n^c \right)^c$.

• Similarly $\bigcup_{i=1}^{\infty} A_i \in \mathcal{L} \Leftarrow (2), (1), \text{ 1st half}$

Def: Let X be an abstract set

(1) \mathcal{F} on X is a collection of $\subset P(X)$ s.t.:

(a) $\emptyset \in \mathcal{F}$. (b) $A \in \mathcal{F} \Rightarrow A^c \in \mathcal{F}$. (c) $A_1, A_2, \dots \in \mathcal{F} \Rightarrow \bigcup_{i=1}^{\infty} A_i \in \mathcal{F}$.

(2) If \mathcal{F} is a σ -algebra. We call (X, \mathcal{F}) a measurable space

Prop. Suppose (X, \mathcal{F}) is a measurable space. Then

- (1) If $A_1, A_2 \in \mathcal{F}$, then $A_1 \cup A_2, A_1 \cap A_2, A_1 \setminus A_2, A_1 \Delta A_2 \in \mathcal{F}$.
- (2) If $A_1, A_2, \dots \in \mathcal{F}$, then $\bigcap_{n=1}^{\infty} A_n \in \mathcal{F}$.
- (3) $X \in \mathcal{F}$.

Rmk: (1) We can replace (c) by (c'): $\bigcap_{n=1}^{\infty} A_n \in \mathcal{F}$.

(2) If we replace (c) by a weaker condition

$$(c') A_1, \dots, A_n \in \mathcal{L} \Rightarrow \bigvee_{i=1}^n A_i \in \mathcal{F}$$

Then the resulting \mathcal{F} is called a Boolean algebra.

Example 1 $X = \mathbb{R}^d$. $\mathcal{F} = \mathcal{L}$. (Lebesgue algebra).

2. (Null algebra) $X = \mathbb{R}^d$. $\mathcal{F} = \{A \in \mathcal{L} : m(A) = 0 \text{ or } m(A^c) = 0\}$.

3. (Trivial algebra) $\mathcal{F} = \{\emptyset, X\}$.

4. (Discrete algebra) $\mathcal{F} = P(X)$

5. (Atomic algebra) $X = \bigvee_{\alpha \in I} A_\alpha$. \leftarrow disjoint union.

$$\rightarrow \mathcal{F} = \left\{ \bigvee_{\alpha \in J} A_\alpha : J \subseteq I \right\}.$$

6. (Elementary algebra) $X = \mathbb{R}^d$

$$\mathcal{F} = \{E \subset \mathbb{R}^d : E \text{ is elementary or } E^c \text{ is elementary}\}$$

Then \mathcal{F} is a Boolean algebra on X .
but not a σ -algebra.

Similarly: Jordan algebra.

Prop. Let $\{\mathcal{F}_\alpha\}$ be a family of σ -algebras on X .

Then $\mathcal{F} := \bigcup \mathcal{F}_\alpha$ is a σ -algebra on X .

Proof: • $\emptyset \in \mathcal{F}_\alpha \Rightarrow \emptyset \in \mathcal{F}$.

• $A \in \mathcal{F} \Rightarrow A \in \mathcal{F}_\alpha \forall \alpha \Rightarrow A^c \in \mathcal{F}_\alpha \Rightarrow A^c \in \mathcal{F}$.

• $A_n \in \mathcal{F} \Rightarrow A_n \in \mathcal{F}_\alpha \forall \alpha \Rightarrow \bigcup_{n=1}^{\infty} A_n \in \mathcal{F}_\alpha \Rightarrow \bigcup_{n=1}^{\infty} A_n \in \mathcal{F}$.

Def. Let $\mathcal{X} \subset P(X)$ be any collection of subsets of X . We define
 $\sigma(\mathcal{X}) = \langle \mathcal{X} \rangle =$ the intersection of all σ -algebra that contains \mathcal{X}
 $(= \text{the smallest } \sigma\text{-algebra that contains } \mathcal{X})$.

We call $\sigma(\mathcal{X})$ the σ -algebra generated by \mathcal{X} .

Example Let $X = \mathbb{R}^d$

$$\begin{aligned} X &= \text{all open sets in } \mathbb{R}^d && \text{Borel algebra} \\ \rightarrow \langle X \rangle &= \text{all Borel sets in } \mathbb{R}^d = \overset{\uparrow}{B} \neq L \\ &\quad \downarrow \sigma\text{-algebra generated by all open sets on } \mathbb{R}^d \end{aligned}$$

Rmk. Recall: $A \in L \Leftrightarrow A = F \cup N$, where F is F_σ , N is null.

The Lebesgue algebra is generated by the Borel algebra and the Null algebra.

2. The measure space

Recall: The Lebesgue measure $m: [L \rightarrow [0, +\infty]]$ satisfies

(1) $m(\emptyset) = 0$

(2) If A_1, A_2, \dots disjoint, then $m(\bigcup_{n=1}^{\infty} A_n) = \sum_{n=1}^{\infty} m(A_n)$

(3) For any A_1, A_2, \dots , we have $m(\bigcup_{n=1}^{\infty} A_n) \leq \sum_{n=1}^{\infty} m(A_n)$

(4) If $A_1 \subset A_2 \subset \dots$ then $m(\bigcup_{n=1}^{\infty} A_n) = \lim_{n \rightarrow \infty} m(A_n)$ from m sub-additivity

(5) If $A_1 \supset A_2 \supset \dots$, $m(A_1) < +\infty$, then $m(\bigcap_{n=1}^{\infty} A_n) = \lim_{n \rightarrow \infty} m(A_n)$:

(6) $m(A + \{x\}) = m(A)$ ← not define on abstract set

(7) $A_1 \subset A_2 \Rightarrow m(A_1) \leq m(A_2)$

(8) (Fatou) If $A_1, A_2, \dots \in L$, then $m(\liminf_{n \rightarrow \infty} A_n) \leq \liminf_{n \rightarrow \infty} m(A_n)$.

(9) If $A_1, A_2, \dots \in L$, $m(\bigcup_{n=1}^{\infty} A_n) < \infty$, then $m(\limsup_{n \rightarrow \infty} A_n) \geq \limsup_{n \rightarrow \infty} m(A_n)$

(10) (Dominant convergence). Suppose $|A_n| \rightarrow A$.

$(X_{A_n} \rightarrow X_A \text{ pointwise})$

$\exists F \in L$, $F \supset A_n$ for all n , $m(F) < \infty$, then $m(A) = \lim_{n \rightarrow \infty} m(A_n)$

(11) If $m(A) = 0$, $B \subset A$, then $m(B) = 0$

$\boxed{ABC A}$

$\boxed{B \subset A}$

Def let (X, \mathcal{F}) be a measurable space

1) A map $\mu: \mathcal{F} \rightarrow [0, +\infty]$ is called a measure if

$$(a) \mu(\emptyset) = 0$$

$$(b) \mu(\bigcup_{n=1}^{\infty} A_n) = \sum_{n=1}^{\infty} \mu(A_n) \text{ for disjoint } A_n \in \mathcal{F}.$$

2) We call (X, \mathcal{F}, μ) a measure space

Prop: || Let (X, \mathcal{F}, μ) be a measurable space

then (3), (4), (5), (7), (8), (9), (10) hold.

Rmk • We say (X, \mathcal{F}, μ) a finite measure space if $\mu(X) < \infty$.

• We say (X, \mathcal{F}, μ) a σ -finite measure space if

$$\exists A_n \in \mathcal{F} \text{ s.t. } \mu(A_n) < \infty \text{ s.t. } X = \bigcup_{n=1}^{\infty} A_n.$$

• We say (X, \mathcal{F}, μ) a probability measure space if $\mu(X) = 1$

Example 1 (Lebesgue measure) $X = \mathbb{R}^d$, $\mathcal{F} = \mathcal{L}$, $\mu = m$.

2. For any non-negative measurable function f on \mathbb{R}^d , one can define a measure $\mu_f: \mathcal{L} \rightarrow [0, +\infty]$
by $\mu_f(A) := \int_A f d\lambda$.

3. (Dirac measure) Fix $x_0 \in X$, we define

$$\delta_{x_0}: \mathcal{L} \rightarrow [0, +\infty], \text{ by } \delta_{x_0}(A) = \chi_A(x_0) = \begin{cases} 1 & x_0 \in A \\ 0 & x_0 \notin A \end{cases}$$

Example 4. (The counting measure) For any $A \in \mathcal{F}$.

$$\# A = \begin{cases} \text{number of elements in } A, \text{ if } A \text{ is a finite set} \\ \infty \quad \text{if } A \text{ is an infinite set} \end{cases}$$

More generally, For any nonnegative function f on X , one can define $\#_f A = \sup_{(a_n) \subset A} \sum_n f(a_n)$
(for A is an uncountable set).

5. Suppose (X, \mathcal{F}, μ) is a measure space $\overset{Y \in X}{\rightarrow}$
 $\rightarrow \mathcal{F}_Y = \{A \cap Y : A \in \mathcal{F}\}$ is a σ -algebra on Y .
 If $Y \in \mathcal{F}$, then $\mu_Y(A \cap Y) := \mu(A \cap Y)$ is a measure
 on (Y, \mathcal{F}_Y) .

6. Suppose μ_1, μ_2, \dots are measures on (X, \mathcal{F})
 Then for any $c_1, c_2, \dots \geq 0$, the function $\mu := \sum_{n=1}^{\infty} c_n \mu_n / \sum c_n$
 is still a measure on (X, \mathcal{F}) .

* Property III) doesn't hold for abstract measure space (X, \mathcal{F}, μ) .

Suppose $A \in \mathcal{F}$, and $\mu(A) = 0$.

Q: Do we have $\mu(B) = 0$ for $\forall B \subseteq A$? (i.e. $B \in \mathcal{F}$)

Problem: It's possible that $A \in \mathcal{F}$, $\mu(A) = 0$, $B \subseteq A$ but $B \notin \mathcal{F}$.
 e.g. Consider Dirac measurable on $(\mathbb{R}^d, \mathcal{F}_{\Delta})$

$$\forall A \in \mathbb{Z}, x_0 \notin A \Rightarrow \delta_{x_0}(A) = 0.$$

But there exists non-measurable set \tilde{A} that doesn't
 contain x_0 . Since $\tilde{A} \notin \mathbb{Z}$, so we don't have $\delta_{x_0}(\tilde{A}) = 0$.

Defn|| We say a measure space (X, \mathcal{F}, μ) is complete if it
 satisfies $\mu(A) = 0$, $\underbrace{B \subseteq A}_{\text{(on } X)}$ $\Rightarrow B \in \mathcal{F}$ ($\Rightarrow \mu(B) = 0$).

Fact: Borel algebra is not complete

\exists measure zero Borel set A s.t. $\exists B \subseteq A$, B is not Borel.

$$\rightarrow (X, \mathcal{B}, \mu)$$

Prop: Let (X, \mathcal{F}, μ) be a measure space.

$$N = \{N \in \mathcal{F} : \mu(N) = 0\}.$$

Let $\bar{\mathcal{F}} = \{A \cup B : A \in \mathcal{F}, \text{ and } \exists N \in N \text{ s.t. } B \subset N\}$.

Then $\bar{\mathcal{F}}$ is a σ -algebra on X .. and $\bar{\mu} : \bar{\mathcal{F}} \rightarrow [0, +\infty]$..

defined by $\bar{\mu}(A \cup B) = \mu(A)$
 is a complete measure on $(X, \bar{\mathcal{F}})$

Lee 16

1 Measurable function

Def: A function $f: X \rightarrow R$ (or $[-\infty, +\infty]$) is $(\mathcal{F}\text{-measurable})$.

$\Leftrightarrow \forall t \in R$ the set $\{x: f(x) > t\} \in \mathcal{F}$.

$\Leftrightarrow \forall \epsilon > 0$ the set $\{x: |f(x)| < \epsilon\} \in \mathcal{F}$.

$\Leftrightarrow \forall \text{ open set } V \subset R$ the set $f^{-1}(V) \in \mathcal{F}$.

Closed-Borel

Rank: Let $(X, \mathcal{F}), (Y, \mathcal{G})$ be measurable space

A map $\varphi: X \rightarrow Y$ is called measurable if
 $\varphi^{-1}(B) \in \mathcal{F}$, for any $B \in \mathcal{G}$.

So, A function $f: X \rightarrow R$ is measurable means the map f

$f: (X, \mathcal{F}) \rightarrow (R, \mathcal{B})$ is measurable
 $\uparrow \text{(map)} \quad \uparrow \text{Borel } \sigma\text{-algebra}$

In particular, a function $f: R^d \rightarrow R$ is measurable in Lebesgue theory

means, f is a measurable map $f: (R^d, \mathcal{L}) \rightarrow (R, \mathcal{B})$

Not $f: (R^d, \mathcal{L}) \rightarrow (R, \mathcal{L})$ is measurable. ($\exists A \in \mathcal{L}, \forall x \in f^{-1}(A) \notin \mathcal{L}$)

Prop: Let (X, \mathcal{F}) be a measurable space

(1) $A \in \mathcal{F} \Leftrightarrow \chi_A$ is measurable function

(2) If f_1, f_2 are measurable, then $c_1 f_1 + c_2 f_2, f_1 \circ f_2$ are measurable

(3) If $\varphi: R \rightarrow R$ is continuous, $f: X \rightarrow R$ is measurable.
 then $\varphi \circ f$ is measurable.

(4) f_n are measurable $\Rightarrow \limsup f_n, \liminf f_n, \sup f_n, \inf f_n$
 is measurable

Simple function: $f(x) = \sum_{i=1}^n c_i \chi_{A_i}$ $A_i \in \mathcal{F}$

Prop: || Let $f: X \rightarrow [0, +\infty]$ measurable. Then there exists $0 \leq g_1 \leq g_2 \dots \nearrow f$, where each g_i is a simple function on (X, \mathcal{F}) (Moreover, the coverage is uniform on any set $\{x \in A\}$ where f is bounded).
Set where f is bounded \rightarrow a set A converges uniformly.

Proof: $f_n(x) = \begin{cases} \frac{k-1}{2^n} & \text{if } \frac{k-1}{2^n} \leq f(x) \leq \frac{k}{2^n}, k=1, 2, \dots, n^2 \\ n & \text{if } f(x) > n \end{cases}$

MCT Let (X, \mathcal{F}, μ) be a measure space. And
 $0 \leq f_1 \leq f_2 \leq \dots$

Then $\int_X \lim_{n \rightarrow \infty} f_n d\mu = \lim_{n \rightarrow \infty} \int_X f_n d\mu$
→ Linearity, Countability I: $\int_X \sum_{n=1}^{\infty} f_n d\mu = \sum_{n=1}^{\infty} \int_X f_n d\mu$
Corollaries of MCT II: $\int_X \sum_{n=1}^{\infty} \chi_{A_n} f d\mu = \sum_{n=1}^{\infty} \int_{A_n} f d\mu$
 $A_n \in \mathcal{F}$, disjoint

2. Theory of integration: Non-negative case.

△ Let $g(x) = \sum_{i=1}^n c_i \chi_{A_i}$ be a non-negative simple function
 $\int_X g d\mu = \sum_{i=1}^n c_i \mu(A_i)$

Again, need to check that the integral is independent of the representation of g .

△ Let (X, \mathcal{F}, μ) be a measure space

Let $f: X \rightarrow [0, +\infty]$ be a non-negative measurable function

Def: || $\int_X f d\mu := \sup \{ \int_X g d\mu : 0 \leq g \leq f, g \text{ simple} \}$

Countable additivity III Let (X, \mathcal{F}, μ) be a measurable space
 $f: X \rightarrow [0, \infty]$ be a non-negative measurable function. Then for any sequence μ_1, μ_2, \dots of measure on \mathcal{F} .

$$\int_X f d(\sum_{n=1}^{\infty} \mu_n) = \sum_{n=1}^{\infty} \int_X f d\mu_n$$

Proof. - $f = \chi_A$ $A \in \mathcal{F}$ LHS = $\sum_{n=1}^{\infty} \mu_n(A) = \text{RHS}$

- Linearity \rightarrow simple function
- MCT \rightarrow all functions ≥ 0 .

Example. Let $X = N \{1, 2, 3, \dots\}$
 $\mathcal{F} = P(X)$
 $\mu = \#$ "counting measure"

Note: Any sequence of numbers a_1, a_2, a_3, \dots can be viewed as a function $a: N \rightarrow \mathbb{R}$, $a(n) := a_n$.

It's \mathcal{F} -measurable

Suppose $a_n \geq 0$ for all n . Then

$$\int_X a d\mu = \lim_{N \rightarrow \infty} \int_X a^N d\mu = \lim_{N \rightarrow \infty} \sum_{n=1}^N a(n) = \sum_{n=1}^{\infty} a_n$$

$$a^N(n) = \begin{cases} a(n) & n \in N \\ 0 & n > N \end{cases}$$

"Infinite Series = the integral over N w.r.t. the counting measure"

Countable additivity I: View f_1, f_2, \dots as one function on

$$f: X \times N \rightarrow [0, \infty] \quad f(x, n) := f_n(x)$$

Then $\int_X (\sum_n f(x, n) d\#_n) d\mu_x = \sum_n \int_X f(x, n) d\mu_x$
 Tonelli: $(\int_X \sum_n f_n d\mu_x = \sum_n \int_X f_n d\mu_x)$

3. Integration: absolute convergence theory

Def || Let (X, \mathcal{F}, μ) be a measure space, f a measurable function

We say f is absolutely integrable if

$$\|f\|_{L^1(X, \mathcal{F}, \mu)} := \int_X |f| d\mu < \infty$$

For such f , define

$$\int_X f d\mu = \int_X f^+ d\mu - \int_X f^- d\mu.$$

$\rightarrow L^1(X, \mathcal{F}, \mu)$ or $L^P(X, \mathcal{F}, \mu)$

$$\text{e.g. } L^1(N, P(X), \#) = \{(\alpha_n) : \sum_{n=1}^{\infty} |\alpha_n| < +\infty\}$$

$$L^P(N, P(X), \#) = \{(\alpha_n) : \sum_{n=1}^{\infty} |\alpha_n|^p < +\infty\}$$

\rightarrow Monotonicity, Linearity, countable additivity. Absolute Continuity

Given $f \in L^1(X, \mathcal{F}, \mu)$ $\forall \epsilon > 0 \exists s$ s.t.
 $\forall A: \mu(A) < s. \int_A |f| d\mu < \epsilon$

D.C.T: $f_n \rightarrow f$ a.e. or in measure $\nrightarrow f \in L^1(X, \mathcal{F}, \mu)$
 $\exists g \in L^1(X, \mathcal{F}, \mu)$ s.t. $|f_n| \leq g$ and $f_n \rightarrow f$ in $L^1(X, \mathcal{F}, \mu)$ norm
 $\Leftrightarrow \int_X |f_n - f| d\mu \rightarrow \int_X g d\mu$

\rightarrow Completeness of L^1 . (L^P)

What could be different?

$(X, \mathcal{F}), (Y, \mathcal{G})$ be measurable spaces

$\rightarrow (X \times Y, \mathcal{F} \otimes \mathcal{G})$

the σ -algebra generated by sets of the form $A \times B$.
 $A \in \mathcal{F}, B \in \mathcal{G}$.

Fact | Suppose $(X, \mathcal{F}, \mu), (Y, \mathcal{G}, \nu)$ be two measure spaces.

Then \exists a unique measure $\mu \times \nu$.

on $\mathcal{F} \otimes \mathcal{G}$ s.t:

$$\mu \times \nu(A \times B) = \mu(A) \cdot \nu(B).$$

σ -finite
i.e. $X = \bigcup_{n=1}^{\infty} X_n$
 $\mu(X_n) < \infty$

Tonelli Thm (Incomplete Version)

Let $(X, \mathcal{F}, \mu), (Y, \mathcal{G}, \nu)$ be two σ -finite measure spaces

$f: X \times Y \rightarrow [0, +\infty]$ be a $\mathcal{F} \otimes \mathcal{G}$ measurable function. Then

(1) $f(\cdot, y)$ is measurable on X ($\forall y$ on Y).

(2) $y \mapsto \int_X f(x, y) dx$ is measurable on Y .

(3) $\int_{X \times Y} f(x, y) d\mu d\nu = \int_Y (\int_X f(x, y) d\mu) d\nu = \int_X (\int_Y f(x, y) d\nu) dx$.

Example: $\Delta X = [0, 1] \quad F = P(X) \quad \mu = \# \rightarrow \text{NOT } \sigma\text{-finite}$
and

$\Delta Y = [0, 1] \quad F = L \quad \mu = m$

$$f(x, y) = \begin{cases} 1 & x = y \\ 0 & x \neq y \end{cases}$$

$$\int_Y (\int_X f(x, y) d\#_X) dm_Y = \int_Y 1 dm = 1$$

$$\int_X (\int_Y f(x, y) dm_Y) d\#_X = \int_X 0 d\# = 0$$

Difference: $L^{k_1+k_2} \neq L^{k_1} \otimes L^{k_2}$

FACT $L^{k_1+k_2} \neq L^{k_1} \otimes L^{k_2}$

But $B^{k_1+k_2} = B^{k_1} \otimes B^{k_2}$

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Recall: Elementary \rightarrow Jordan $\xrightarrow{\sim}$ Lebesgue outer measure m^*

$\xrightarrow{\sim}$ Lebesgue measure m

1. Outer measure:

The difference between m^* and m .

- m^* is defined for all sets.

m is defined for $A \in L$ and

- Subadditive (countable)

$m(A) = m^*(A)$

$$m^*(\bigcup_{n=1}^{\infty} A_n) \leq \sum_{n=1}^{\infty} m^*(A_n)$$

Additive (countable)

$$m(\bigcup_{n=1}^{\infty} A_n) = \sum_{n=1}^{\infty} m(A_n) \text{ for disjoint } A_n's$$

Def: For any set X . An outer measure is a map $\mu^*: P(X) \rightarrow [0, +\infty]$
s.t.

(1) $\mu^*(\emptyset) = 0$.

(2) If $A_1 \subset A_2$, then $\mu^*(A_1) \leq \mu^*(A_2)$

(monotonicity)

• (3) $\mu^*(\bigcup_{n=1}^{\infty} A_n) \leq \sum_{n=1}^{\infty} \mu^*(A_n)$

(countable subadditivity)

Recall: For abstract measure, additivity + non-negative \rightarrow monotonicity
 For outer measure, subadditivity + non-negative \rightarrow monotonicity

To construct measure and σ -algebra:

① Construct outer measure \rightarrow ② Construct σ -algebra and ③ Construct outer measure on it

Prop. Let \mathcal{E} be any collection of subsets of X . (i.e. $\mathcal{E} \subset P(X)$).
 Such that $\bigvee_{E \in \mathcal{E}} E = X$, $\emptyset \in \mathcal{E}$.

Let $\varphi: \mathcal{E} \rightarrow [0, +\infty]$ be any function s.t. $\varphi(\emptyset) = 0$.

For any $A \subset X$. let (cause contradiction?)

$$\mu^*(A) = \inf \left\{ \sum_{n=1}^{\infty} \varphi(E_n) \mid A \subset \bigcup_{n=1}^{\infty} E_n, E_n \in \mathcal{E} \right\}.$$

(If A can't be covered by countable E_n 's, then def $\mu^*(A) = +\infty$)

Then $\mu^*: P(X) \rightarrow [0, +\infty]$ is an outer measure.

Proof: • Obviously $\mu^*(\emptyset) = 0$, $\mu^*(A_1) \leq \mu^*(A_2)$ if $A_1 \subset A_2$.

• For countable additivity: If $\exists A$ s.t. $\mu^*(A) = +\infty$,
 then $\mu^*(\bigcup_{n=1}^{\infty} A_n) \leq \sum_{n=1}^{\infty} \mu^*(A_n)$.

Next suppose each $\mu^*(A_n) < +\infty$.

$$\begin{aligned} \text{By def, } \exists E_m^{(n)} \in \mathcal{E} \text{ s.t. } A_n \subset \bigcup_{m=1}^{\infty} E_m^{(n)} \text{ and} \\ \sum_{m=1}^{\infty} \varphi(E_m^{(n)}) \leq \mu^*(A_n) + \varepsilon \cdot \frac{1}{2^n}. \end{aligned}$$

$$\Rightarrow \bigcup_{n=1}^{\infty} A_n \subset \bigcup_{n=1}^{\infty} \bigcup_{m=1}^{\infty} E_m^{(n)}$$

$$\begin{aligned} \Rightarrow \mu^*\left(\bigcup_{n=1}^{\infty} A_n\right) &\leq \mu^*\left(\bigcup_{n,m=1}^{\infty} E_m^{(n)}\right) \leq \sum_{n=1}^{\infty} \mu^*(A_n) + \varepsilon. (\forall \varepsilon > 0). \\ \Rightarrow \varepsilon \rightarrow 0 \quad \mu^*\left(\bigcup_{n=1}^{\infty} A_n\right) &\leq \sum_{n=1}^{\infty} \mu^*(A_n) \end{aligned}$$

From outer measure \rightarrow measure [i.e. find \mathcal{F}].

Recall: $A \in \mathcal{L} \iff \forall \varepsilon > 0 \exists \text{ open set } V > A \text{ s.t. } m^*(V \setminus A) < \varepsilon$
 $\iff \forall \varepsilon > 0 \exists \text{ closed set } F \subset A \text{ s.t. } m^*(A \setminus F) < \varepsilon$

$$\begin{aligned}
 &\Leftrightarrow A = G \setminus N, G = G_S, N = \text{null set} \\
 &\Leftrightarrow A = F \cup N, F = F_O, N = \text{null set} \\
 \text{Carathéodory} \Leftrightarrow &\forall \text{ box } B \quad |B| : m^*(B \cap A) + m^*(B \setminus A). \leftarrow \begin{array}{l} \text{If } A \subset B \text{ (bounded)} \\ \text{it's enough to check} \\ \text{this for } B. \end{array} \\
 &\forall T \subset \mathbb{R}^d \quad m^*(T) = m^*(T \cap A) + m^*(T \setminus A).
 \end{aligned}$$

Def: // Let μ^* be an outer measure on a set X

We say a subset $A \subset X$ is (Carathéodory) measurable if

$$\mu^*(T) = \mu^*(T \cap A) + \frac{\mu^*(T \setminus A)}{\mu^*(T \setminus A)} \quad \forall T \subset X. \quad (*)$$

Rank: By subadditivity we already has

$$\mu^*(T) \leq \mu^*(T \cap A) + \mu^*(T \setminus A).$$

So it's enough to check \geq

Lemma: // If $\mu^*(A) = 0$, then A is measurable.

Proof: $\mu^*(T) \geq \mu^*(T \cap A) + \mu^*(T \setminus A)$

Then (Carathéodory extension thm) Suppose $\mu^*: P(X) \rightarrow [0, +\infty]$ is any outer measure, let $\mathcal{F} = \{A \subset X : A \text{ satisfies } (*)\}$. Then $(1) \mathcal{F}$ is a σ -algebra, and $(2) \mu = \mu^*|_{\mathcal{F}}$ is a measure.

Proof. (1) $\emptyset \in \mathcal{F}$. In fact, it's a complete measure.

$$A \in \mathcal{F} \Rightarrow \mu^*(T) = \mu^*(T \cap A) + \mu^*(T \cap A^c) \Rightarrow A^c \in \mathcal{F}.$$

Now suppose $A_1, A_2, \dots \subset \mathcal{F}$.

- First prove $A_1 \cup A_2 \in \mathcal{F}$:

$$\begin{aligned}
 \mu^*(T) &= \mu^*(T \cap A_1) + \mu^*(T \cap A_1^c) \\
 &= \underbrace{\mu^*(T \cap A_1) \cap A_2}_{\uparrow T \cap (A_1 \cup A_2)} + \underbrace{\mu^*(T \cap A_1^c) \cap A_2^c}_{T \cap (A_1 \cap A_2^c)} \\
 &\quad + \underbrace{\mu^*(T \cap A_2) \cap A_1^c}_{T \cap (A_1 \cap A_2)} + \underbrace{\mu^*(T \cap A_2^c) \cap A_1}_{T \cap (A_1^c \cap A_2)} \\
 &\geq \mu^*(T \cap (A_1 \cup A_2)) + \mu^*(T \cap A_1^c \cap A_2^c)
 \end{aligned}$$

- $\bigcup_{n=1}^{\infty} A_n \in \mathcal{F}$.

- Write $B = \bigcup_{n=1}^{\infty} A_n$. $B_N = \bigcup_{n=1}^N A_n \in \mathcal{F}$.

NOTE: $\mu^*(T \cap B_N^c) \geq \mu^*(T \cap B^c)$

$$\boxed{\text{WANT: } \mu^*(T) \geq \mu^*(T \cap B) + \mu^*(T \cap B^c)}$$

$$\rightarrow \lim_{N \rightarrow \infty} \mu^*(T \cap B_N^c) \geq \mu^*(T \cap B^c).$$

$$\mu^*(T \cap B_{N+1}) = \mu^*(T \cap B_N) + \mu^*(T \cap (B_{N+1} \setminus B_N))$$

$$\Rightarrow \lim_{N \rightarrow \infty} \mu^*(T \cap B_{N+1}) = \sum_{n=1}^{\infty} \mu^*(T \cap (B_{n+1} \setminus B_n))$$

$$\Rightarrow \mu^*(T) \geq \sum_{n=1}^{\infty} \mu^*(T \cap (B_{n+1} \setminus B_n)) + \mu^*(T \cap B^c)$$

$$\geq \mu^*(T \cap B) + \mu^*(T \cap B^c)$$

(2) $\mu(\emptyset) = \mu^*(\emptyset) = 0$.

- Let $A_n \in \mathcal{F}$ disjoint. WANT: $\mu^*\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} \mu^*(A_n)$

$$\mu^*(A_1 \cup A_2) = \mu^*(A_1 \cup A_2 \cap A_2) + \mu^*(A_1 \cup A_2 \cap A_2^c) = \mu^*(A_1) + \mu^*(A_2)$$

$$\mu^*(A_1 \cup A_2 \cup \dots \cup A_N) = \sum_{n=1}^N \mu^*(A_n)$$

$$\mu^*\left(\bigcup_{n=1}^{\infty} A_n\right) \geq \mu^*\left(\bigcup_{n=1}^N A_n\right) = \sum_{n=1}^N \mu^*(A_n) \xrightarrow{N \rightarrow \infty} \mu^*\left(\bigcup_{n=1}^{\infty} A_n\right) \geq \sum_{n=1}^{\infty} \mu^*(A_n)$$

$$\Rightarrow \mu^*\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} \mu^*(A_n).$$

2. Premeasure \leftarrow How to construct a "nice" outer measure μ^*

Recall: Elementary or Jordan \rightsquigarrow Lebesgue

Q: finite additivity \nrightarrow Countable additivity
on a Boolean algebra \mathcal{B} on a σ -algebra \mathcal{F}

A. In general, no.

Example: $X = N$ $\mathcal{B} = P(X)$ $\mu_0(A) = \begin{cases} 0 & \text{if } A \text{ is finite} \\ \infty & \text{if } A \text{ is not finite} \end{cases}$
 $\mathcal{F} = P(X)$ finite additivity

$$\mu_0(N) = 0 + \sum_{n=1}^{\infty} \mu_0(\{n\}).$$

Def: A premeasure is a finite additive measure μ_0 on a Boolean algebra \mathcal{B} . i.e. $A \in \mathcal{B} \Leftrightarrow \bigcup_{n=1}^{\infty} A_n$ For disjoint A_n , $\mu_0\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} \mu_0(A_n)$. which satisfies an additional property.

If A_1, A_2, \dots disjoint, and if $\bigcup_{n=1}^{\infty} A_n \in \mathcal{B}$.
then $\mu_0\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} \mu_0(A_n)$.

Example: Let (X, \mathcal{F}, μ_X) and (Y, \mathcal{G}, μ_Y) be measure spaces.

Let $\mathcal{B} = \{ \bigcup_{i=1}^k (A_i \times B_i) \cup (A_i \times B'_i) \cup \dots \cup (A_k \times B_k) : A_i \in \mathcal{F}, B_i \in \mathcal{G} \}$.

Can check: \mathcal{B} is a Boolean algebra. Can suppose $A_i \times B_i$'s are disjoint

Define: $\mu_0\left(\bigcup_{i=1}^k (A_i \times B_i)\right) := \sum_{i=1}^k \mu_X(A_i) \times \mu_Y(B_i)$.

One can check that μ_0 is a premeasure.

e.g. A simple case $A \times B = \bigcup_{n=1}^{\infty} (A_n \times B_n)$ $A_n \times B_n$'s are disjoint

$$\Rightarrow \chi_A(x) \chi_B(y) = \chi_{A \times B} = \sum_{n=1}^{\infty} \chi_{A_n \times B_n} = \sum_{n=1}^{\infty} \chi_{A_n}(x) \cdot \chi_{B_n}(y)$$

$$\Rightarrow \int \chi_{A \times B} d\mu_X = \sum_{n=1}^{\infty} \mu_X(A_n) \cdot \chi_{B_n}(y)$$

$$\Rightarrow \mu_X(A) \mu_Y(B) = \sum_{n=1}^{\infty} \mu_X(A_n) \cdot \mu_Y(B_n).$$

Thm (Hahn-Kolmogorov) Let \mathcal{B} be a Boolean algebra on X .

μ_0 is a premeasure on \mathcal{B} . Then there exists a σ -algebra $\mathcal{F} \supset \mathcal{B}$ and a measure μ on \mathcal{F} s.t. $\mu = \mu_0$ for $A \in \mathcal{B}$. i.e. (\mathcal{F}, μ) extends (\mathcal{B}, μ_0)

\rightarrow We get $\underline{\mu_X \times \mu_Y}$ on $X \times Y$.

Proof: We have seen

$\mu^*(A) := \inf \left\{ \sum_{n=1}^{\infty} \mu_0(E_n) : E_n \in \mathcal{B}, A \subset \bigcup_{n=1}^{\infty} E_n \right\}$.

= outer measure on X .

$\mathcal{F} = \{A \subset X : \mu^*(T) = \mu^*(T \cap A) + \mu^*(T \cap A^c), \forall T \subset X\} \rightarrow \sigma\text{-algebra}$

remains to check:

$\mu = \mu^*$ is a measure

1) $B \subset T$

2) For $A \in B$, we have $\mu^*(A) = \mu_0(A)$.

3) Let $A \in B$. Need $\mu^*(T) \geq \mu^*(T \cap A) + \mu^*(T \setminus A)$.

If $\mu^*(T) = +\infty$ ✓.

Now assume $\mu^*(T) < +\infty \Rightarrow \exists E_n \in B, T \subset \bigcup_{n=1}^{\infty} E_n$.

$$\sum_{n=1}^{\infty} \mu_0(E_n) \leq \mu^*(T) + \varepsilon.$$

$$T \cap A \subset \bigcup_{n=1}^{\infty} E_n \cap A \Rightarrow \mu^*(T \cap A) \leq \sum_{n=1}^{\infty} \mu^*(E_n \cap A) \leq \sum_{n=1}^{\infty} \mu_0(E_n \cap A).$$

$$\mu^*(T \setminus A) = \mu^*(T \cap A^c) \leq \sum_{n=1}^{\infty} \mu_0(E_n \cap A^c)$$

$$\mu^*(T \setminus A) + \mu^*(T \cap A) \leq \sum_{n=1}^{\infty} (\mu_0(E_n \cap A) + \mu_0(E_n \cap A^c)) = \sum_{n=1}^{\infty} \mu_0(E_n) \leq \mu^*(T) + \varepsilon.$$

2) By def. only need $\mu^*(A) \geq \mu_0(A)$.

Suppose $A \subset \bigcup_{n=1}^{\infty} E_n, E_n \in B, \mu^*(A) \geq \sum_{n=1}^{\infty} \mu_0(E_n) - \varepsilon$. $A \subset \bigcup_{n=1}^{\infty} \tilde{E}_n$
Let $\tilde{E}_n = \left(\bigcup_{i=1}^n E_i \setminus \bigcup_{i=1}^{n-1} E_i \right) = \tilde{E}_n \setminus \bigcup_{i=1}^{n-1} E_i$ Then \tilde{E}_n disjoint $= \bigcup_{n=1}^{\infty} \tilde{E}_n$.

$$\Rightarrow \mu_0(A) = \mu_0\left(\bigcup_{n=1}^{\infty} (\tilde{E}_n \cap A)\right) = \sum_{n=1}^{\infty} \mu_0(\tilde{E}_n \cap A) \leq \sum_{n=1}^{\infty} \mu(\tilde{E}_n) \leq \sum_{n=1}^{\infty} \mu(E_n) \leq \mu^*(A) + \varepsilon.$$

The extension μ of μ_0 is unique when μ is σ -finite
(on \mathcal{M})

Lec 18 Metric v.s. measure

1. Metric Outer measure

In \mathbb{R}^d , we have open sets, closed sets, compact sets.

They all come from the = metric structure

the distance between two points.

Def: Let X be a set. A distance function d is a function

$$d: X \times X \rightarrow [0, +\infty)$$

$$\text{s.t. (1) } d(x, y) = d(y, x)$$

$$(2) d = 0 \Leftrightarrow x = y$$

$$(3) d(x, y) + d(y, z) \geq d(x, z) \quad \forall x, y, z \in X. \text{ (Triangle inequality)}$$

We say (X, d) is a metric space.

Example: $X = \mathbb{R}^d$. $d(x, y) = \|x - y\| = \sqrt{\sum_{i=1}^d (x_i - y_i)^2}$

Example: X any set $d(x, y) = \begin{cases} 1 & x \neq y \\ 0 & x = y \end{cases}$

Example: $X = L^1(\mathbb{R}^d)$. (If $f = g$ a.e. then we regard f and g as the same element in $L^1(\mathbb{R}^d)$)

$$d(f, g) = \int_X |f - g| dx.$$

Given a metric space (X, d) , we can define

- Open ball: $B_r(x) = \{y \in X. d(x, y) < r\}$.

- Open set U : $\forall x \in U \exists r > 0$. s.t: $B_r(x) \subset U$.

- Closed set F : $\Leftrightarrow F^c$ is open.

- Compact set K : $\Leftrightarrow K$ is closed and bounded. not hold for example 2.

\Leftrightarrow For $K \subset \bigcup U_\alpha$. U_α is open. one can find finitely many U_1, \dots, U_N s.t: $K \subset \bigcup_{n=1}^N U_n$.

- \mathcal{F}_{σ} : = countable union for closed sets.
- \mathcal{G}_δ : = countable intersection for open sets
- Borel set: the σ -algebra generated by all open sets (or closed sets) (\mathcal{B}_X).

Def: Any measure μ on \mathcal{B}_X is called Borel measure.

e.g. $m|_{\mathcal{B}}$ is a Borel measure. ($\Leftrightarrow \mathcal{L} \supset \mathcal{B}$)
 (Lebesgue)

To prove this: • \mathcal{L} is a σ -algebra.

- \mathcal{L} contains all open/closed sets.

We used (Pset2 - Part 1) $d(A, B) > 0 \Rightarrow m^*(A \cup B) = m^*(A) + m^*(B)$

Def: We say an outer measure μ^* on a metric space (X, d) is a metric outer measure if

$$d(A, B) > 0 \Rightarrow \mu^*(A \cup B) = \mu^*(A) + \mu^*(B)$$

Rmk: $A, B \subset X$.

$$d(A, B) = \inf \{d(x, y) : x \in A, y \in B\}.$$

This is NOT a distance function on $P(X)$.

because $A \cap B \neq \emptyset \Rightarrow d(A, B) = 0$.

Thm // Let μ^* be a metric outer measure on (X, d)

// Then all Borel sets are measurable w.r.t. μ^* .

→ the induced measure $\mu^*|_{\mathcal{B}}$ is a Borel measure

f. Proof: $\mu^* \rightarrow \mathcal{F}$ σ -algebra of all measurable sets.
 Want: $\mathcal{B}_X \subset \mathcal{F}$.

So: It's enough to prove: all closed sets $\bar{T} \in \mathcal{F}$.

$$\text{i.e.: } \mu^*(T) = \mu^*(T \cap \bar{T}) + \mu^*(T \setminus \bar{T}).$$

W.L.O.G. we assume $\mu^*(T) < \infty$.

For each n . let

$$T_n = \{x \in T \setminus F : d(x, F) \geq \frac{1}{n}\} \subset T \setminus F. \quad d(x, F) = 0 \text{ if } x \in F$$

$$\text{Then } T \setminus F = \bigcup_{n=1}^{\infty} T_n.$$

Since μ^* is metric outer measure, $\mu^*(T) \geq \mu^*((T \cap F) \cup T_n)$.

$$= \mu^*(T \cap F) + \mu^*(T_n).$$

$$\text{Fact: } \lim_{n \rightarrow \infty} \mu^*(T_n) = \mu^*(T \setminus F).$$

$$\Rightarrow \mu^*(T) \geq \mu^*(T \cap F) + \mu^*(T \setminus F).$$

Prop: Suppose $\mu^*(\mu)$ is a metric outer measure. Borel measure

$$\text{Suppose } \mu(B_\delta(r)) < \infty. \quad \forall x \in X, \forall r \in \mathbb{R}.$$

(Then \forall Borel set A . $\forall \epsilon > 0$. \exists open $U \supset A$. closed $F \subset A$.

s.t: $\mu(U \setminus F) < \epsilon \Rightarrow \mu(U \setminus A) < \epsilon, \mu(A \setminus F) < \epsilon$) Regularity

2. Hausdorff measure

$$r^d, \|x-y\|.$$

Given any $A \subset \mathbb{R}^d$ bounded. we define the diameter

$$\text{diam}(A) = \sup \{d(x, y), x, y \in A\}.$$

Def: For any $\alpha \geq 0$. $\delta > 0$. define

$$h_{\alpha, \delta}^*(A) = \inf \left\{ \sum_{k=1}^{\infty} (\text{diam } A_k)^{\alpha} : A \subset \bigcup_{k=1}^{\infty} A_k, \text{diam}(A_k) < \delta \right\}. \quad \text{by } \mathcal{S}(\mathcal{I})$$

and let

$$h_{\alpha}^*(A) = \lim_{\delta \rightarrow 0} h_{\alpha, \delta}^*(A)$$

Prop: \mathbb{H}_{α} is a metric outer measure on \mathbb{R}^d

\rightarrow We get a Borel measure $\mathbb{H}_{\alpha}^*(A) = h_{\alpha}^*(A) \quad A \in \mathcal{B}$.

α -dimensional Hausdorff measure.

Proof: $h_\alpha^*(\phi) = 0$

$$\cdot A_1 \subset A_2 \Rightarrow h_\alpha^*(A_1) \leq h_\alpha^*(A_2).$$

$$\cdot A_1, A_2, \dots \Rightarrow h_\alpha^*(\bigcup_{n=1}^{\infty} A_n) \leq \sum_{n=1}^{\infty} h_\alpha^*(A_n).$$

It's enough to prove $h_{\alpha, \delta}^*(\bigcup_{j=1}^{\infty} A_j) \leq \sum_{j=1}^{\infty} h_{\alpha, \delta}^*(A_j)$. ($\forall \delta$).

Take $F_{j,k}$ with $\text{diam}(F_{j,k}) < \delta$.

$$A_j \subset \bigcup_{k=1}^{\infty} F_{j,k}.$$

$$\sum_{k=1}^{\infty} \text{diam}(F_{j,k})^\alpha \leq h_{\alpha, \delta}^*(A_j) + \frac{\delta}{2j}.$$

$$\Rightarrow \bigcup_{j=1}^{\infty} A_j \subset \bigcup_{j,k} F_{j,k}.$$

$$\Rightarrow h_{\alpha, \delta}^*(\bigcup_{j=1}^{\infty} A_j) \leq \sum_{j=1}^{\infty} h_{\alpha, \delta}^*(A_j) + \varepsilon. \quad (\text{let } \varepsilon \rightarrow 0, \delta \rightarrow 0).$$

$$\text{Want: } h_\alpha^*(A_1 \cup A_2) = h_\alpha^*(A_1) + h_\alpha^*(A_2)$$

Now let $d(A_1, A_2) > 0$.

Take $\delta < d(A_1, A_2)$

$$\text{Let } F_1, F_2, \dots \text{ such that } \text{diam}(F_i) < \delta. \quad A_1 \cup A_2 \subset \bigcup_{i=1}^{\infty} F_i$$

Let F'_1, F'_2, \dots be those that intersects A_1 .

$$F'_1, F'_2, \dots \subset A_1 \cup A_2.$$

$$\text{Then } \{F'_1, F'_2, \dots, F''_1, F''_2, \dots\} \subset \{F_1, F_2, \dots\}.$$

$$\Rightarrow \sum_j \text{diam}(F'_j)^\alpha + \sum_i \text{diam}(F''_i)^\alpha \leq \sum_k \text{diam}(F_k)^\alpha.$$

$$\Rightarrow h_\alpha^*(A_1 \cup A_2) \geq h_\alpha^*(A_1) + h_\alpha^*(A_2)$$

$$\text{Therefore } h_\alpha^*(A_1 \cup A_2) = h_\alpha^*(A_1) + h_\alpha^*(A_2)$$

Prop. 11) If $H^\alpha(A) < \infty$, then $\forall \beta > \alpha$, $H^\beta(A) = 0$.

(2) If $H^\beta(A) > 0$, then $\forall \alpha < \beta$, $H^\alpha(A) = +\infty$.

Proof: Suppose $\alpha < \beta$, $\text{diam } F < \delta$.

$$\Rightarrow (\text{diam } F)^\beta = (\text{diam } F)^{\beta-\alpha} (\text{diam } F)^\alpha < \delta^{\beta-\alpha} (\text{diam } F)^\alpha.$$

$$\delta \rightarrow 0$$

Def: For any Borel set A , the Hausdorff dimension of A .

$$\text{is } \dim_H A = \sup \{\beta : H^\beta(A) = \infty\}.$$

$$= \inf \{\alpha : H^\alpha(A) = 0\}.$$

Lee 19 Riesz representation theorem

Recall Lebesgue outer measure satisfies

$$\cdot \mu(A) = \inf \{ \mu(U) : U \supset A, U \text{ open} \}. \quad (\text{i})$$

→ outer regularity

$$\cdot \mu(A) = \sup \{ \mu(K) : K \subset A, K \text{ compact} \} \quad (\text{ii})$$

Regular Measure: (for Borel Measure).

Def. Let: $\mu: \mathcal{B}_X \rightarrow [0, +\infty]$ be a Borel measure.

We call μ is ① outer regular if it satisfies (i) for all $A \in \mathcal{B}_X$

② inner regular - - - - - (ii) - - - - -

③ regular if it is both outer and inner regular

About Compact Sets:

Now Recall: A set K is compact. If for any open covering $\{U_\alpha\}$ of K . (i.e. each U_α is open and $K \subset \bigcup U_\alpha$)

there exists a finite subcovering $\{U_1, \dots, U_N\}$

(i.e. $U_i \in \{U_\alpha\}$ $1 \leq i \leq N$. and $K \subset \bigcup_{i=1}^N U_i$)

Example: Let $K \subset \mathbb{R}^d$ be a bounded closed set

Then K is compact

e.g. $[0, 1]^d$.

Fact 1 If K is compact, FCK is closed, then F is compact

Proof: Take $\{U_\alpha\}$ an open covering of F .

Consider $\{U_\alpha, F^c\}$. is an open covering of K .

$\Rightarrow \exists \{U_1, \dots, U_N, F^c\}$ covers $K \Rightarrow$ covers F .

$\Rightarrow \{U_1, \dots, U_N\}$ covers F .

Fact 2 If f is a continuous function, K is compact, then \exists

$x_0, y_0 \in K$ s.t. $f(x_0) = \inf_{K} \{f(x) : x \in K\}$, $f(y_0) = \sup_{K} \{f(x) : x \in K\}$.

Proof. Suppose NOT. $\Rightarrow \exists \{U_n\} \subset K$. $f(x_1) > f(x_2) > \dots$ & $\inf \{f(x)\}$.
 Let $U_K = \{x : f(x) > f(x_K)\}$. Then $\{U_k\}$ cover K .
 And $U_1 \not\subseteq U_2 \not\subseteq \dots$
 \Rightarrow No finite subcovering

Prop: Let (X, d) be a compact metric space. metric outer measure
Let $\mu : B_X \rightarrow [0, +\infty]$ be a (Borel measure)
Suppose $\mu(X) < +\infty$.
Then μ is regular

Proof. By last problem set up problem 1 (P Set 9 Part 2).
 • $\forall \varepsilon > 0. \exists U$ s.t. $\mu(U \setminus A) < \varepsilon \Rightarrow \mu(U) < \mu(A) + \varepsilon$ for $\forall A \in B_X$.
 • $\exists F$ closed, $F \subset X$ compact $\Rightarrow F$ compact
 s.t. $\mu(A \setminus F) < \varepsilon. \mu(A) < \mu(F) + \varepsilon$.

Thm (Partition of Unity Simplex version).

Let (X, d) be a compact (metric space). Let U_1, \dots, U_N be open sets in X . $K \subset U_1 \cup \dots \cup U_N$ is a compact set. Then \exists continuous functions f_1, \dots, f_N on X s.t.

- ① $f_i(x) \geq 0. \quad \forall i \forall x.$
- ② $\sum_{i=1}^N f_i(x) \leq 1. \quad \forall x \in X.$ $\xrightarrow{\text{Compact in condition of thm.}}$
- ③ $\text{Supp}(f_i) \subset U_i \quad (\text{Supp}(f) = \{x : f(x) \neq 0\})$
- ④ $\sum_{i=1}^N f_i(x) = 1. \quad \forall x \in K.$ $\xrightarrow{\text{closed and } \subset X \text{ (compact)}}$

Proof of Lemma

Proof: Write $U = \bigcup_{i=1}^N U_i$ open $\Rightarrow U^c$ is closed thus compact
 $\Rightarrow d(K, U^c) = \delta > 0.$

$$\begin{aligned} d(K, U^c) &= \inf \{d(x, y) : x \in K, y \in U^c\} \\ &= \inf \{d(x, U^c) : x \in K\}. \end{aligned}$$

Lemma: A X compact metric space. $(U \subset X)$ is open. $K \subset U$ compact. $\exists V$ open s.t. $K \subset V \subset \bar{V} \subset U$.

Consider $g(x) = d(x, U^c)$. Then g is continuous on K .

Let $\mathcal{V} = \{V_i : V_i \text{ open } \exists i \text{ s.t. } \bar{V}_i \subset U_i\}$.

By lemma A \mathcal{V} is an open covering of $K \Rightarrow K = \bigcap_{i=1}^n V_i \cup \dots \cup V_n$.

Let $K_i = \bigcap_{j \in I_i} \bar{V}_j$. Then $K_i \subset U_i$ compact and $\bigcup_{i=1}^n K_i \supset K$.

$$\Rightarrow d(K, U^c) = \inf \{d(x, U^c) : x \in K\} = \underline{d(x_0, U^c)} > 0 \quad (\text{only need } U^c \text{ is closed})$$

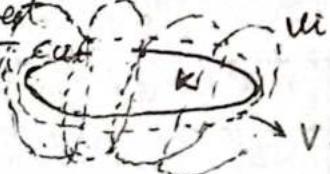
$$\text{Let } V = B(K, \frac{\delta}{2}) = \bigcup_{x \in K} B(x, \frac{\delta}{2})$$

Then V is open, $K \subset V \subset \bar{V} \subset U$

[Let $V_i = U_i \cap V$. Then V_i open, $K \subset V_1 \cup \dots \cup V_n$]

\uparrow in order to ensure ③ But not correct

$$\boxed{f_i(x) = \frac{d(x, W_i)}{d(x, K) + \sum_{i=1}^n d(x, W_i)}}$$



Let

$$\sum f_i(x) = \frac{\sum d(x, V_i^c)}{d(x, K) + \sum d(x, V_i^c)} \quad (0 < \delta)$$

$$x \in K \Rightarrow d(x, K) = 0 \Rightarrow \sum f_i = 1.$$

for ③: $f_i(x) = 0 \text{ if } x \in V_i^c$

$$\Rightarrow \text{Supp}(f_i) \subset \overline{V_i} \subset U_i.$$

Now suppose (X, d) is a compact metric space.

$\mu: B_X \rightarrow [0, +\infty]$ is a Borel measure, $\int \mu(x) < +\infty$.

For any $f \in C(X) \Rightarrow f$ attains max/min values
 $\Rightarrow \exists M \text{ s.t. } |f(x)| \leq M$.

Define $I_\mu(f) = \int_X f d\mu \in \mathbb{R}$.

$$I_\mu(f) = \int_X f d\mu \in \mathbb{R}.$$

We have (1) $I_\mu(c_1 f_1 + c_2 f_2) = c_1 I_\mu(f_1) + c_2 I_\mu(f_2)$.

(2) If $f \geq 0$, then $I_\mu(f) \geq 0$.

Where $I_\mu: C(X) \rightarrow \mathbb{R}$.

Def: We call $\ell: C(X) \rightarrow \mathbb{R}$ is a linear functional if $\ell(c_1 f_1 + c_2 f_2) = c_1 \ell(f_1) + c_2 \ell(f_2)$

We say linear functional ℓ is positive if $f \geq 0 \Rightarrow \ell(f) \geq 0$.

Thm (Riesz representation thm) (Simplest version)

Let (X, d) be a locally compact & compact metric space, and ℓ be a positive linear functional on $C(X)$. Then \exists a unique Borel measure μ s.t: $\ell(f) = \int_X f d\mu$, $\forall f \in C(X)$. regular locally finite
Moreover, $\mu(X) < +\infty$.

Rank: we have seen that such μ must be a regular Borel measure.

Idea: { we need a metric outer measure

we need an outer measure

Suppose we
Get μ : $\mu(U) = \int_X \chi_U d\mu + \ell(\chi_U)$.

But we can find $f_n \uparrow \chi_U$. $f_n(x) = \begin{cases} 0 & \text{if } x \notin U \\ \frac{1}{n} & \text{if } x \in U, d(x, U^c) < \frac{1}{n} \\ 1 & \text{if } x \in U, d(x, U^c) \geq \frac{1}{n}. \end{cases}$
Then $0 \leq f_n \leq 1$. $\text{supp } f_n \subset U$.

Then $\mu(U) = \int_X \chi_U d\mu = \lim_{n \rightarrow \infty} \int_X f_n d\mu = \lim_{n \rightarrow \infty} \ell(f_n)$

Proof. Step 1 (Uniqueness). We have seen that if μ_1, μ_2 are two Borel measures satisfy the theorem, then for any open set U . $\mu_1(U) = \lim_{n \rightarrow \infty} \ell(f_n) = \mu_2(U)$.

By outer regularity

$$\mu_1(A) = \mu_2(A) \quad \forall A \in \mathcal{B}_X.$$

Step 2 For any open $U \subset X$. let

$$\mu^*(U) = \sup \{ \ell(f) : 0 \leq f \leq 1, \text{supp } f \subset U \}.$$

and for any $A \subset X$. let

$$\mu^*(A) = \inf \{ \mu^*(U) : U \supset A \text{ (open)} \}.$$

Claim: μ^* is an outer measure

$$\cdot \mu^*(\emptyset) = 0 \quad \checkmark$$

$$\cdot A_1 \cap A_2 \Rightarrow \mu^*(A_1) = \mu^*(A_2) \quad \checkmark$$

$$\cdot \mu^*(\bigcup_{i=1}^{\infty} A_i) \leq \sum_{i=1}^{\infty} \mu^*(A_i) \dots$$

First. \rightarrow If U, V open then $\mu^*(UVV) \leq \mu^*(U) + \mu^*(V)$.

For any $f: 0 \leq f \leq 1, K = \text{supp}(f) \subset U \cap V$.

By P.O.U. $\Rightarrow \varphi, \psi$ s.t. $\varphi + \psi = 1$ on K .

$0 \leq \varphi, \psi \leq 1, \text{supp}(\varphi) \subset U, \text{supp}(\psi) \subset V$.

$$\Rightarrow f = (\varphi + \psi)f = \varphi f + \psi f \quad 0 \leq \varphi f \leq 1 \quad \text{Supp}(\varphi f) \subset U$$

$$\Rightarrow \ell(f) = \ell(\varphi f + \psi f) \quad 0 \leq \psi f \leq 1 \quad \text{Supp}(\psi f) \subset V.$$

$$\leq \mu^*(U) + \mu^*(V)$$

$$\Rightarrow \mu^*(UVV) \leq \mu^*(U) + \mu^*(V).$$

Then \rightarrow countable additivity for $\{A_n\}$ (WLOG. $\forall A_n, \mu^*(A_n) < \infty$).

Take open $U_n > A_n$ s.t. $\mu^*(U_n) \leq \mu^*(A_n) + \frac{\epsilon}{2^n}$

Let $U = \bigcup_{n=1}^{\infty} U_n, \forall f \in C(X), 0 \leq f \leq 1, \text{supp}(f) = K \subset U$.

By compactness $\exists U_1, \dots, U_N$ s.t. $K \subset \bigcup_{i=1}^N U_i$.

$$\Rightarrow \ell(f) \leq \mu^*\left(\bigcup_{n=1}^{\infty} U_n\right) \leq \sum_{n=1}^N \mu^*(U_n) \leq \sum_{n=1}^N \mu^*(A_n) + \epsilon.$$

$$\leq \sum_{n=1}^{\infty} \mu^*(A_n) + \epsilon.$$

$$\Rightarrow \mu^*\left(\bigcup_{n=1}^{\infty} A_n\right) \leq \mu^*\left(\bigcup_{n=1}^{\infty} U_n\right) \leq \sup \ell(f) \leq \sum_{n=1}^{\infty} \mu^*(A_n).$$

Step 3. μ^* is metric outer measure.

Assume $d(A, B) > 0$. WANT: $\mu^*(A \cup B) = \mu^*(A) + \mu^*(B)$.

Take $V > A \cup B$ s.t. $\mu^*(V) \leq \mu^*(A \cup B) + \epsilon$.

Let $U_1 = U \cap B(C_A, \frac{r}{3}), U_2 = U \cap B(C_B, \frac{r}{3})$.

$\Rightarrow U_1, U_2$ open $V > U_1 \cup U_2 > A \cup B$. $U_1 \cap U_2 = \emptyset$ $\text{d}(U_1, U_2) > \frac{r}{3}$.

Now suppose $f_i \in C(X), 0 \leq f_i \leq 1, \text{supp}(f_i) \subset U_i$.

$$\ell(f_i) \geq \mu^*(U_i) - \epsilon.$$

Let $f = f_1 + f_2$ Then $0 \leq f \leq 1, \text{supp}(f) \subset U_1 \cup U_2$.

$$\Rightarrow \mu^*(A \cup B) \geq \mu^*(V) - \epsilon \geq \mu^*(U_1 \cup U_2) - \epsilon$$

$$\geq \ell(f) - \epsilon = \ell(f_1) + \ell(f_2) - \epsilon$$

$$\geq \mu^*(U_1) + \mu^*(U_2) - 3\epsilon.$$

$$\geq \mu^*(A) + \mu^*(B) - 3\epsilon$$

$\Rightarrow \mu^*$ is metric outer measure $\Rightarrow \mu^*/\emptyset$ is a Borel measure.

Step 4 X is open $\forall 0 < \varepsilon \leq 1$. $\int_X (1-f) \geq 0 \Rightarrow \ell(1) \geq \ell(f)$.
 $\mu^*(X) \leq \ell(1) < +\infty$.

$$(L = \mathbb{Z}\mu)$$

Step 5 $\ell(f) = \int_X f d\mu$.

the proof works for more general version of Riesz repth

$\ell(f) \leq \int_X f d\mu$. If $f(-f) \leq \int_X (-f) d\mu \Rightarrow \ell(f) \geq \int_X f d\mu$.
 $f \in C(X) \Rightarrow a_0 = \inf f, b_0 = \sup f$. on $K = \text{supp}(f)$.

Take $a = y_0 < y_1 < y_2 \dots < y_{n-1} < y_n = b$ s.t. $y_i - y_{i-1} < \varepsilon$

Let $E_i = \{x : y_i < f(x) \leq y_{i+1}\} \in \mathcal{B}_X$. $E_i \cap E_j = \emptyset$.

$$K = \bigcup_{i=1}^n E_i$$

$\exists \tilde{U}_1, \dots, \tilde{U}_n$ open s.t. $\tilde{U}_i \supset E_i$. $\mu(\tilde{U}_i) < \mu(E_i) + \frac{\varepsilon}{n}$

Let $U_i = \tilde{U}_i \cap f^{-1}(-\infty, y_i + \varepsilon)$. Then U_i open. $E_i \subset U_i$, $\mu(U_i) < \mu(E_i) + \frac{\varepsilon}{n}$, $\sup_{U_i} f \leq y_i$

$\bigcup_{i=1}^n U_i = K \Rightarrow \exists \phi_1, \dots, \phi_n$ s.t. $\phi_i \geq 0$, $\text{supp}(\phi_i) \subset U_i$

$$\sum_{i=1}^n \phi_i = 1 \text{ on } K \Rightarrow f = f \cdot \sum_{i=1}^n \phi_i = \sum_{i=1}^n f \cdot \phi_i$$

and $\phi_i f \leq \phi_i(y_i + \varepsilon)$ $\forall i = 1, \dots, n$.

Note: $\sum_{i=1}^n \phi_i = 1$ on $K \Rightarrow \mu(K) \leq \ell(\sum_{i=1}^n \phi_i) = \sum_{i=1}^n \ell(\phi_i)$

This implies

$$\ell(f) = \sum_{i=1}^n \ell(\phi_i f) \leq \sum_{i=1}^n (y_i + \varepsilon) \ell(\phi_i)$$

$$= \sum_{i=1}^n (y_i + |a| + \varepsilon) \ell(\phi_i) - |a| \sum_{i=1}^n \ell(\phi_i)$$

$$\leq \sum_{i=1}^n (y_i + |a| + \varepsilon) \mu(U_i) - |a| \mu(K)$$

$$\leq \sum_{i=1}^n (y_i + |a| + \varepsilon) (\mu(E_i) + \frac{\varepsilon}{n}) - |a| \mu(K) \quad \mu(K) = \sum_{i=1}^n \mu(E_i)$$

$$= \sum_{i=1}^n (y_i + \varepsilon) \mu(E_i) + (y_i + |a| + \varepsilon) \frac{\varepsilon}{n}$$

$$\leq \sum_{i=1}^n (y_i - \varepsilon) \mu(E_i) + \varepsilon (2\mu(K) + b + |a| + \varepsilon)$$

$$\leq \int_X f d\mu + \varepsilon (2\mu(K) + b + |a| + \varepsilon). \quad \text{Let } \varepsilon \rightarrow 0. \quad \ell(f) \leq \int_X f d\mu = \mathbb{Z}\mu f$$

\angle^* Space

Lee 20

1. Spaces and Structures

- Spaces \Rightarrow a set contains objects that we want to study
- there are "structures" that describes relations among those objects
- We have learned the following structures on "Space V" (elements v, w, \dots)

- (1) Linear structure: $v \rightarrow av$ $v, w \rightarrow v+w$ $\left. \begin{array}{l} \text{linear space} \\ (\text{vector space}) \end{array} \right\}$ e.g. \mathbb{R}^d , $L(\mathbb{R}^d)$, $C_c(X)$
- (2) Multiplication structure: $v, w \rightarrow v \cdot w \in V$ (inner product is not a multiplication) e.g. $\mathbb{R}, \mathbb{R}^3, (\vec{v}, \vec{w} \rightarrow \vec{v} \times \vec{w})$, $C_c(X)$.
(x. convolution) ...

- (3) Norm structure: A function $\| \cdot \|: V \rightarrow \mathbb{R}$, s.t.

$$\left\{ \begin{array}{l} (1) \|0\|=0 \quad \|v\|>0 \text{ for } v \neq 0. \\ (2) \|av\|=|a|\|v\|. \\ (3) \|v+w\| \leq \|v\| + \|w\| \end{array} \right. \quad \begin{array}{l} \text{vector space (Requires)} \\ \|x\| \end{array}$$

e.g. $V = \mathbb{R}^d, \sqrt{x_1^2 + \dots + x_d^2}$
 $\mathbb{R}^d, |x_1| + \dots + |x_d|$
 $\mathbb{R}^d, \sup(|x_1|, \dots, |x_n|)$
 $L^1(\mathbb{R}^d), \int_{\mathbb{R}^d} |f(x)| dx$.

- (4) Metric structure: A function $d: X \times X \rightarrow \mathbb{R}$, s.t.
(Not require X to be a vector space)

$$\left\{ \begin{array}{l} (1) d(x, x)=0, \quad d(x, y)>0 \text{ for } x \neq y \\ (2) d(x, y)=d(y, x) \\ (3) d(x, z) \leq d(x, y) + d(y, z) \end{array} \right.$$

Note: Norm structure induces a metric structure by
 $d(v, w) = \|v-w\|$.

- (5) Topological structure: "define open sets". $\mathcal{T} \subset \mathcal{P}(X)$.

- $\emptyset \in \mathcal{F}$
 - $x_1, \dots, x_n \in \mathcal{F} \Rightarrow \bigcap_{i=1}^n x_i \in \mathcal{F}$ (finite)
 - $x_\alpha \in \mathcal{F} \Rightarrow \bigcup_\alpha x_\alpha \in \mathcal{F}$.
- $\left. \begin{array}{l} \text{closed set} \\ \text{compact set (by covering)} \\ \text{convergence} \\ \text{continuous function} \end{array} \right\}$

Note: For any metric space. one can study the "metric topology"
i.e. the topology generated by open balls

For Space of functions, one of the very important aspect
in topology is "how functions converges".

(6) Inner Product on Vector Space $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}$ (or \mathbb{C}).

$$(1) \quad \langle av_1 + bv_2, w \rangle = a\langle v_1, w \rangle + b\langle v_2, w \rangle.$$

$$(2) \quad \langle v, w \rangle = \langle w, v \rangle.$$

$$(3) \quad \langle 0, 0 \rangle = 0. \quad \langle v, v \rangle > 0. \quad \forall v \neq 0.$$

Note: $\langle \cdot, \cdot \rangle \rightarrow \|v\| = \sqrt{\langle v, v \rangle}$

we have "angles": $\angle(v, w) = \arccos \frac{\langle v, w \rangle}{\|v\| \cdot \|w\|}$

- (7) Measure Structure.
- (8) Order Structure
- (9) Symmetry
- (10) Duality

Given V . find another V^* s.t. $\forall \ell \in V^* \quad \forall v \in V$.

"pairing" $\langle \ell, v \rangle \in \mathbb{R}$.

e.g. inner product function.

Measure \leftrightarrow function. $\langle u, f \rangle := \int_X f d\mu$

linear functional \leftrightarrow function. $\langle \ell, f \rangle := \ell(f)$.

e.g. Riesz rep. thm

2. L^p Spaces

- Let $X \subset \mathbb{R}^d$ measurable, or more generally (X, \mathcal{F}, μ) .
be a measure space.

Def. (1) For any measurable function f , the L^p -norm

$$\text{of } f \text{ is } \|f\|_{L^p(X)} := \left(\int_X |f|^p d\mu \right)^{\frac{1}{p}}$$

(2) The space of L^p functions

$$= L^p(X) = \{f: f \text{ is measurable } \|f\|_{L^p} < \infty\}$$

Example: $L^p(\mathbb{R}^d)$

Example: $X = \mathbb{N}$, $\mathcal{F} = P(\mathbb{N})$, $\mu = \#\text{-counting measure}$

Then • a function \mapsto a sequence $(a) = (a_1, a_2, \dots)$

• Any function is measurable

$$L^p\text{-norm: } \|a\|_{L^p} = \left(\sum_{i=1}^{\infty} |a_i|^p \# \right)^{\frac{1}{p}} = \left(\sum_{i=1}^{\infty} |a_i|^p \right)^{\frac{1}{p}}$$

$$L^p\text{-space } \ell^p = \{a: \left(\sum |a_i|^p \right)^{\frac{1}{p}} < \infty\}$$

Rank: We always regard a.e. equal functions as the same function. In other words, L^p -space is

$$L^p(X) = \{f: f \text{ measurable, } \|f\|_{L^p} < \infty\} / \sim$$

$$\sim: f_1 \sim f_2 \Leftrightarrow f_1 = f_2 \text{ a.e.}$$

L^p -Vector Space

Prop: II For each $0 < p < \infty$, $L^p(X)$ is a vector space

Proof: $\|af\|_p < \infty \Rightarrow \|af\|_p = \left(\int_X |af|^p d\mu \right)^{\frac{1}{p}} = |a| \|f\|_p < \infty$

$$\text{i.e. } f \in L^p \Rightarrow a \cdot f \in L^p.$$

We need $f, g \in L^p \Rightarrow f+g \in L^p$

$$(f+g)^p \leq (2 \max(|f|, |g|))^p \leq 2^p (|f|^p + |g|^p)$$

$$\Rightarrow \|f+g\|_p^p \leq 2^p (\|f\|_p^p + \|g\|_p^p) < \infty$$

$$\|f+g\|_p < \infty$$

L^p ($p \geq 1$) — normed space

OUR STORY BEGINS

$p \geq 1$

Thm: The L^p -norm $\|\cdot\|_{L^p}$ is a norm on $L^p(X)$.

i.e.: 1) $\|0\|_{L^p} = 0$, $\|f\|_{L^p} > 0$ for $\forall f \neq 0$

2) $\|af\|_{L^p} = |a| \|f\|_{L^p}$

3) $\|f+g\|_{L^p} \leq \|f\|_{L^p} + \|g\|_{L^p}$ (Minkowski's inequality) } Triangle inequality
 $\|\cdot\|_{L^p}$ is a convex function on L^p

→ Metric structure → topological structure → Convergence in L^p -norm

Lemma: If $f \in L^p$, $p > 0$. $\|f^s\|_{L^p} = \|f\|_{L^{ps}}$

Proof: RHS = $(\int_X |f|^s d\mu)^{\frac{1}{ps}} = \int_X |f|^p d\mu)^{\frac{1}{p}} = \|f\|_{L^p} = \text{LHS}$.

Thm (Hölder's inequality) // Suppose $p, q > 0$, $r > 0$, $\frac{1}{p} + \frac{1}{q} = \frac{1}{r}$.

Suppose $f \in L^p$, $g \in L^q$

Then $f, g \in L^r$ and $\|fg\|_{L^r} \leq \|f\|_{L^p} \|g\|_{L^q}$.

Proof: WLOG. assume $f, g \geq 0$, $A = \|f\|_{L^p} > 0$, $B = \|g\|_{L^q} > 0$.

First assume: $r = 1$, $(\frac{1}{p} + \frac{1}{q} = 1)$

$\frac{1}{p} + \frac{1}{q} = 1$ (P>1)

need: (Young's inequality). $a, b \geq 0 \Rightarrow ab \leq \frac{1}{p}a^p + \frac{1}{q}b^q$

$$\Rightarrow \frac{\|f \cdot g\|_{L^1}}{AB} = \int_X \frac{f}{A} \cdot \frac{g}{B} d\mu \leq \int_X \frac{1}{p} \left(\frac{f}{A}\right)^p + \frac{1}{q} \left(\frac{g}{B}\right)^q d\mu$$

$$= \frac{1}{pA^p} \int_X f^p d\mu + \frac{1}{qB^q} \int_X g^q d\mu = \frac{1}{p} + \frac{1}{q} = 1.$$

For general r . $\|fg\|_{L^r} = \|(fg)^{\frac{1}{r}}\|_{L^r} = \|f^{\frac{1}{r}} g^{\frac{1}{r}}\|_{L^1}$
 $\leq \|f^{\frac{1}{r}}\|_{L^p} \|g^{\frac{1}{r}}\|_{L^q} = \|f\|_{L^p} \|g\|_{L^q}$.

Proof: Note $p = 1 \vee$. WLOG $f, g \geq 0$.

$(P>1)$, Let $\frac{1}{p} + \frac{1}{q} = 1 \Rightarrow p+q = pq \Rightarrow p = q(p-1)$

$$\|f+g\|_{L^p}^p = \int_X f(f+g)^{p-1} d\mu + \int_X g(f+g)^{p-1} d\mu.$$

$$\leq \|f\|_{L^p} \cdot \|f+g\|_{L^p} + \|g\|_{L^p} \cdot \|f+g\|_{L^p}^{p-1}.$$

$$= (\|f\|_{L^p} + \|g\|_{L^q}) \|f+g\|_{L^{q(p-1)}}^p.$$

$$= (\|f\|_{L^p} + \|g\|_{L^q}) \|f+g\|_{L^p}^p.$$

L^P normed vector space \rightarrow metric space.
Lec 21

• Recall, we have many spaces $L^P(X)$.

Those spaces are not non-intersecting

In last P set If $|f(x)| < \infty$, then

$P_1 < P_2 \Rightarrow L^{P_2}(X) \subset L^{P_1}(X) \leftarrow \| \cdot \|_{L^{P_1}}$ is a norm on $L^{P_2}(X)$.

$\|f\|_{L^P} \rightarrow \|f\|_{L^{P_0}}$ for $P \rightarrow P_0^-$ (\mathcal{Q} : what if $P \rightarrow P_0^+$?)

1. \mathcal{Q} : what happens if $P \rightarrow +\infty$?

Want: $\|f\|_{L^\infty} = \lim_{P \rightarrow \infty} \|f\|_{L^P}$

e.g.: Take $X = [0, 4]$ $f = \begin{cases} 8 & x=0 \\ 4 & 0 < x \leq 1 \\ 1 & 1 < x \leq 3 \\ -2 & 3 < x \leq 4 \end{cases}$

$$\|f\|_{L^P} = \left(\int_{[0,4]} |f|^P dx \right)^{1/P} = (4^P + 2 \cdot 2^P + 1 \cdot 1^P)^{1/P} \xrightarrow{P \rightarrow \infty} 4$$

The maximal value of f ignoring the value of f on a null set
 $P = +\infty$? L^∞ space

[Def:] Let (X, \mathcal{F}, μ) be a measure space, f is a measurable function

- 1) We say f is essentially bounded if $\exists M$ s.t. $|f(x)| \leq M$ a.e. $x \in X$.
- 2) The L^∞ -norm of f is $\|f\|_{L^\infty} = \inf \left\{ M : \sup_{\text{supp}(f(x))} |f(x)| \leq M \text{ a.e. } x \in X \right\}$.
- 3) The L^∞ -space $L^\infty(X) = \{f : \|f\|_{L^\infty} < +\infty\}$.

Thm. $L^\infty(X)$ is a normed vector space. and $\|\cdot\|_{L^\infty}$ is a norm.

Proof. $\forall a \in \mathbb{R}, f \in L^\infty(X) \Rightarrow \exists M$ s.t. $|f| \leq M$ a.e. $\|af\|_{L^\infty} = \|a\| \cdot \|f\|_{L^\infty}$

$$\Rightarrow |af| \leq |a|M \text{ a.e. } \|af\|_{L^\infty} \leq \|a\| M = \|a\| \|f\|_{L^\infty}$$

$\forall f, g \in L^\infty(X) \Rightarrow \exists M_1, M_2$ s.t. $|f| \leq M_1$ a.e. $|g| \leq M_2$ a.e.

$$\Rightarrow |f+g| \leq M_1 + M_2 \text{ a.e. } \|f+g\|_{L^\infty} \leq \|f\|_{L^\infty} + \|g\|_{L^\infty}$$

$\|f\|_{L^\infty} \geq 0$. $\|f\|_{L^\infty} = 0 \Rightarrow f$ a.e. = 0

$$\|f+g\|_{L^\infty} \leq \|f\|_{L^\infty} + \|g\|_{L^\infty}$$

Minkowski For $1 \leq p \leq \infty$ $\|f+g\|_p \leq \|f\|_p + \|g\|_p$.

Hölder For $\frac{1}{p}, \frac{1}{q}, r \leq \infty$ $\frac{1}{p+q} = \frac{1}{r}$ $\Rightarrow \|fg\|_r \leq \|f\|_p \|g\|_q$.

Proof: enough to prove $p=\infty$ $g=r$ case.

$$\|f \cdot g\|_r = \left(\int_X |f \cdot g|^r d\mu \right)^{1/r} \leq \|f\|_{\infty} \left(\int_X |g|^r d\mu \right)^{1/r} = \|f\|_{\infty} \|g\|_r$$

Rmk: One can easily prove Minkowski and Hölder's inequality for more than 2 functions.

$$1 \leq p \leq \infty. \|f_1 + \dots + f_n\|_p \leq \|f_1\|_p + \dots + \|f_n\|_p.$$

$$0 < p_1, \dots, p_n, r \leq \infty \quad \frac{1}{p_1} + \frac{1}{p_2} + \dots + \frac{1}{p_n} = \frac{1}{r} \Rightarrow \|f_1 + \dots + f_n\|_r \leq \|f_1\|_{p_1} + \dots + \|f_n\|_{p_n}$$

Rmk: For $X = \mathbb{N}$, $\mathcal{F} = P(\mathbb{N})$, $\mu = \#\$. $\rightarrow L^{\infty}(X, \mu) = \ell^{\infty}$.

$\ell^{\infty} = \{(a_1, a_2, \dots) : \sup_{n \geq 1} |a_n| < \infty\}$ = the space of all bounded seqs.

$$\|(a_1, a_2, \dots)\|_{\ell^{\infty}} = \sup_{n \geq 1} |a_n|$$

L^p -metric Space

2. L^p -space as metric space

For $1 \leq p \leq \infty$. Define L^p -metric on $L^p(X, \mu)$

$$d_{L^p(X)}(f, g) = \|f - g\|_p.$$

Then (1) $d_{L^p(X)}(f, g) \geq 0 \Leftrightarrow f = g \text{ a.e.} \Leftrightarrow f = g \text{ in } L^p(X)$.

$$(2) d_{L^p(X)}(f, g) = d_{L^p(X)}(g, f)$$

$$(3) d_{L^p(X)}(f, h) \leq d_{L^p(X)}(f, g) + d_{L^p(X)}(g, h) \quad \forall f, g, h \in L^p(X).$$

So $(L^p(X), d_{L^p(X)})$ is a metric space.

Def: We say f_n converges to f in L^p -norm if

$$\lim_{n \rightarrow \infty} \|f_n - f\|_p = 0.$$

Convergence and Completeness

Def: we say a sequence $f_n \in L^p(X)$ is a Cauchy sequence

if $\forall \varepsilon > 0 \exists N$ s.t. $\forall n, m > N. \|f_n - f_m\|_p < \varepsilon$.

Def: A metric space is complete if any Cauchy sequence converges.

Def: A normed space V is called a Banach Space if it's complete.

Thm. (Biesz-Fischer) // For $1 \leq p < +\infty$, $L^p(X)$ is a Banach space.

Proof: (Similar to lecture 12)

Let $\{f_n\}$ is a Cauchy sequence in $L^p(X)$.

We need to find $f \in L^p(X)$ s.t. $\|f_n - f\|_p \rightarrow 0$.

Idea: $f = f_1 + \sum_{i=2}^{\infty} (f_i - f_{i-1})$,

then $f_n \rightarrow f$. converge?

$$g = \|f_1\| + \sum_{i=2}^{\infty} |f_i - f_{i-1}|$$

$$|f_{ni} - f_{n(i-1)}|$$

First assume $1 \leq p < +\infty$.

Step 1. Pick a subsequence f_{n1}, f_{n2}, \dots .

$$\text{s.t. } \|f_{nk} - f_{n(k-1)}\|_p < \frac{1}{2^k}$$

Step 2. Construct g .

$$g_k = \|f_{n1}\| + \sum_{i=2}^k |f_{ni} - f_{n(i-1)}|$$

By monotone $g_k \rightarrow g = \|f_{n1}\| + \sum_{i=2}^{\infty} |f_{ni} - f_{n(i-1)}|$.

$$\Rightarrow g \in L^p(X)$$

$$(\|g_k\|_p \leq \|f_{n1}\|_p + \sum_{i=2}^k \|f_{ni} - f_{n(i-1)}\|_p \leq \|f_{n1}\|_p + 1, < \infty)$$

Moreover, $|f_{nk}| = |f_{n1} + \sum_{i=2}^k (f_{ni} - f_{n(i-1)})| \leq g_k \leq g$.

Step 3: Let $f = f_{n1} + \sum_{i=2}^{\infty} (f_{ni} - f_{n(i-1)}) = \lim_{k \rightarrow \infty} f_{nk}$.

$$\Rightarrow \|f_{nk} - f\|_p \rightarrow 0$$

$$\begin{aligned} |f_{nk}| &\leq g \Rightarrow \|f_{nk} - f\| \leq 2g \Rightarrow f_{nk} \rightarrow f \Rightarrow \\ (\|f_{nk} - f\|_p^p) &\rightarrow 0 \end{aligned}$$

Step 4: Use triangle inequality to show $\|f_n - f\|_p \rightarrow 0$. By LDCT.

Now consider $p = +\infty$.

Step 1. Same. Step 2. $f = f_{n1} + \sum_{i=2}^{\infty} (f_{ni} - f_{n(i-1)})$

$$\|f\|_{+\infty} \leq \|f_{n1}\|_{+\infty} + 1 \text{ a.e. } f \in L^{+\infty}(X)$$

$$\|f_{nk} - f\|_{+\infty} = \left\| \sum_{i=N}^{\infty} (f_{ni} - f_{n(i-1)}) \right\|_{+\infty} \leq \frac{1}{2^{N-1}} \rightarrow 0$$

Step 3. Same..

L^p is Separable.

[Def] Let (V, d) be a metric space.

We say V is separable if there exists a countable set

$$V_0 = \{v_1, v_2, \dots\} \text{ s.t. } \overline{V_0} = V \text{ i.e. } \forall v \in V, \forall \epsilon > 0, \exists i \text{ s.t. } |v_i - v| < \epsilon.$$

e.g. \mathbb{R}^d

[Thm] For $1 \leq p < \infty$, $L^p(\mathbb{R}^d)$ is separable.

[More generally] Suppose X satisfies: (\exists countable many open sets U_1, U_2, \dots s.t. ① $\bigcup U_i$) $\times \mathbb{N}$

② For any open set $U \subset X$, exist U_{n_1}, U_{n_2}, \dots s.t. $U = \bigcup_{i=1}^{\infty} U_{n_i}$.

then $L^p(X, \mu)$ is separable.

NOTE: Not true for $p = \infty$.

Lemma For $0 < p < \infty$, the set

$$\mathcal{S} = \{f \in L^p : f(x) = \sum_{i=1}^k a_i \chi_{A_i} : \mu(A_i) < \infty, k \in \mathbb{N}\} \text{ is dense in } L^p(X).$$

[In particular $(\mathcal{S}, \| \cdot \|_1, \|\cdot\|_p)$ is NOT a Banach Space].

Proof: [Step 1] $B = \{f \in L^p : \exists M \text{ s.t. } |f(x)| \leq M\}$ is dense in L^p .

i.e. $\forall f \in L^p \exists |f_n| \in B$ s.t. $\|f_n - f\|_{L^p} \rightarrow 0$.

Let $f_n = \begin{cases} f(x) & |f(x)| \leq n \\ 0 & |f(x)| > n \end{cases}$ Then $f_n \in B$. $f_n \rightarrow f$.

Use DCT. $|f_n| \leq |f| \Rightarrow |f_n - f| \leq 2|f| \quad \|f_n - f\|_{L^p} \rightarrow 0$.

$$\Rightarrow \|f_n - f\|_{L^p} \rightarrow 0.$$

$$\|f_n - f\|^p \leq 2^p \|f\|^p \in L^1$$

[Step 2] $BB = \{f \in B : \mu(\text{Supp } f) < \infty\}$ is dense in B .

$$\text{Note } \int_X |f|^p d\mu = A < \infty \Rightarrow \mu(\{x : |f(x)| > \frac{1}{m}\}) \leq A m^p$$

Let $f_m = \begin{cases} f(x) & \text{if } |f(x)| > \frac{1}{m} \\ 0 & \text{if } |f(x)| \leq \frac{1}{m} \end{cases}$

$\forall f \in B$

Then $f_m \in BB$, $|f_m| \leq |f|$. $\|f_m - f\|_{L^p} \rightarrow 0$. $\|f_m - f\|^p \leq \|f\|^p$

Step 3. \mathcal{S} is dense in $\mathbb{B}\mathbb{B}$.

Let $f \in BB$. Let $A_i = \{x : \frac{i}{2^k} \leq f(x) < \frac{i+1}{2^k}\}$.

Let $\varphi_K = \sum_{\substack{i \\ \text{finite}}} \frac{i}{2^K} \chi(A_i) \in \mathcal{S}$.
 $\mu(A_i) < \infty$.

$$\| \varphi_k - f \|_{L^p} \rightarrow 0.$$

Step 4. $\forall f \in L^p(X)$ find $f_1 \in \mathcal{B}$ s.t. $\|f_1 - f\|_{L^p} < \frac{\varepsilon}{4}$.

Find $f_2 \in \mathcal{B}\mathcal{B}$ s.t. $\|f_2 - f_1\|_{L^p} < \frac{\epsilon}{4}$. Find $\varphi \in \mathcal{S}$ s.t. $\|\varphi - f_2\|_{L^p} < \frac{\epsilon}{4}$.

Applying Triangle Inequality

Prove: $L^3(X)$ is separable. (in thm's condition)

$$V_0 = \left\{ \sum_{i=1}^n a_i X_{ui}, \quad a_i \in \mathbb{Q}, \quad U_i = V_i, V_i - V_{Uik} \right\}_{k=1}^{ck}.$$

L² Space

Lcc22

L^2 -spaces as inner product spaces

• Consider $p=2$ Then

$$\|f_1 + f_2\|_{L^2} \leq \|f_1\|_{L^2} + \|f_2\|_{L^2}$$

$$\|f_1 \cdot f_2\|_{L^1} \leq \|f_1\|_{L^2} \cdot \|f_2\|_{L^2} \quad (\text{H\"older})$$

$\angle^2(X)$ is complete

• $L^2(\mathbb{R}^n)$ or $\ell^2(\mathbb{Z})$ (= $L^2(N)$, $\mu = \#\#$) are separable.

Recall In \mathbb{R}^d , we have inner product

$$\langle \cdot, \cdot \rangle : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$$

$$\langle \vec{v}, \vec{w} \rangle \rightarrow \sum_{i=1}^d v_i w_i$$

In (12), we can define an inner product

$$\langle \cdot, \cdot \rangle : \ell^2 \times \ell^2 \rightarrow \mathbb{R}$$

$$\langle (a), (b) \rangle = \sum_{i=1}^{\infty} a_i b_i = \int_N a \cdot b d\mu.$$

\uparrow , Hölder \Rightarrow convergence

Similarly, we can define an inner product on $L^2(X)$.

$$\langle \cdot, \cdot \rangle : L^2(X) \times L^2(X) \rightarrow \mathbb{R}$$

$$\langle f, g \rangle := \int_X f \cdot g \, d\mu \cdot <+\infty.$$

Rmk:

$$f, g \in L^2(X) \Rightarrow f \cdot g \in L^1(X) \quad \int_X |f \cdot g| \, d\mu < +\infty.$$

(2) If f, g are complex-valued, then

$$\langle f, g \rangle := \int_X f \cdot \bar{g} \, d\mu.$$

Prop: The inner product $\langle \cdot, \cdot \rangle$ on $L^2(X)$ satisfies.

$$(1) \quad \langle f, f \rangle \geq 0, \text{ and } = 0 \Leftrightarrow \begin{cases} f = 0 \text{ a.e. in } L^2(X), \\ f = 0 \text{ zero function.} \end{cases}$$

$$(2) \quad \langle a_1 f_1 + a_2 f_2, g \rangle = a_1 \langle f_1, g \rangle + a_2 \langle f_2, g \rangle.$$

$$(3) \quad \langle f, g \rangle = \overline{\langle g, f \rangle}.$$

$$(\langle f, g \rangle = \langle g, f \rangle)$$

Def: For any vector space V , if a function $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}$ satisfies (1). (2). (3) then we say $(V, \langle \cdot, \cdot \rangle)$ is an inner product space

Cauchy-Schwarz inequality

$$\text{By Hölder} \quad |\langle f_1, f_2 \rangle| \leq \|f_1\|_{L^2} \cdot \|f_2\|_{L^2} \text{ in } L^2(X).$$

$$(\int_X |f_1 f_2|^2 \, d\mu)^{\frac{1}{2}} \leq (\int_X |f_1|^2 \, d\mu)^{\frac{1}{2}} \cdot (\int_X |f_2|^2 \, d\mu)^{\frac{1}{2}}.$$

$$\text{Special space. } X = \{1, \dots, n\}, \mu = \# \quad \left(\sum_{i=1}^n a_i b_i \right) \leq \left(\sum_{i=1}^n a_i^2 \right)^{\frac{1}{2}} \left(\sum_{i=1}^n b_i^2 \right)^{\frac{1}{2}}$$

Let $(V, \langle \cdot, \cdot \rangle)$ be any inner product space.

Then we define a norm on V by $\|v\| = \sqrt{\langle v, v \rangle}$.

One can check: • $\|\cdot\|$ is a norm on V

• One always has

$$|\langle v, w \rangle| \leq \|v\| \cdot \|w\| \text{ (Cauchy-Schwarz)}$$

Def || If an inner product space $(V, \langle \cdot, \cdot \rangle)$ is complete
 we say $(V, \langle \cdot, \cdot \rangle)$ is a Hilbert Space
 By def. Any Hilbert Space is a Banach Space

Examples: ① \mathbb{R}^d $\boxed{\begin{array}{l} X = \{1, 2, \dots, d\}, \mu = \# \\ \Rightarrow L^2(X) = \mathbb{R}^d \end{array}}$
 ② $L^2(X)$.

Example: $V = C([0,1])$, $\langle \cdot, \cdot \rangle$ in $L^2([0,1])$.

$$\langle f, g \rangle := \int_0^1 f \cdot g \, dx.$$

Then $(V, \langle \cdot, \cdot \rangle)$ is inner product space.

But NOT Complete.

Now let $(V, \langle \cdot, \cdot \rangle)$ be an inner-product space.

$$\begin{aligned} v, w \in V. \Rightarrow \|v+w\|^2 &= \langle v+w, v+w \rangle = \langle v, v \rangle + \langle v, w \rangle + \langle w, v \rangle + \langle w, w \rangle \\ &= \|v\|^2 + \|w\|^2 + 2\langle v, w \rangle. \end{aligned}$$

Def: || The angle between v, w is

$$\theta = \arccos \frac{\langle v, w \rangle}{\|v\| \cdot \|w\|}$$

In particular $v \perp w \Leftrightarrow \langle v, w \rangle = 0$.

Prop: || (Parallelogram rule) In $(V, \langle \cdot, \cdot \rangle)$.

$$\|v+w\|^2 + \|v-w\|^2 = 2(\|v\|^2 + \|w\|^2)$$

$$\text{Proof: } \|v+w\|^2 = \|v\|^2 + \|w\|^2 \pm 2\langle v, w \rangle$$

Can check: In $L^p(X)$ ($p \neq 2$). one can find f, g s.t:

$$\|f+g\|_{L^p}^2 + \|f-g\|_{L^p}^2 \neq 2(\|f\|_{L^p}^2 + \|g\|_{L^p}^2)$$

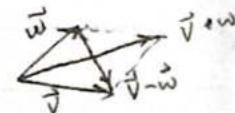
\Rightarrow In L^p ($p \neq 2$) one can't define an inner product

$$\langle \cdot, \cdot \rangle \text{ s.t: } \langle f, f \rangle = \|f\|_p^2.$$

If a norm satisfies: $\|v+w\|^2 + \|v-w\|^2 = 2(\|v\|^2 + \|w\|^2)$

$$\langle v, w \rangle = \frac{1}{4}(\|v+w\|^2 - \|v-w\|^2) \text{ is an inner product.}$$

Then (Existence of minimizer) Let H be a Hilbert space
 $V \cap H$ is a nonempty, closed, convex subset. Then for
 $\forall x \in H$, there exists a unique $y \in V$ s.t:
 $\|x - y\| = \inf \{ \|x - z\| : z \in V\}$.



Proof. (Uniqueness). If y_1, y_2 are two minimizers $y_1 \neq y_2$.
 $y_1, y_2 \in V \Rightarrow (y_1 + y_2)/2 \in V$.

By Parallelogram rule, $\|x - \frac{y_1 + y_2}{2}\|^2 + \|\frac{y_1 - y_2}{2}\|^2 = \|x - y_1\|^2$
 Contradiction.

(Existence) Let $\{z_n\}_{n=1}^{\infty} \subset V$ s.t: $\|x - z_n\| \rightarrow d = \inf \{ \|x - z\| : z \in V\}$.
 Then $(z_n + z_m)/2 \in V$.

$$\Rightarrow d^2 \leq \|x - \frac{z_n + z_m}{2}\|^2 \leq \|x - \frac{z_n + z_m}{2}\|^2 + \left\| \frac{z_n - z_m}{2} \right\|^2 \geq \frac{1}{2} \|x - z_n\|^2 + \frac{1}{2} \|x - z_m\|^2.$$

$$\Rightarrow \|z_n - z_m\|^2 \rightarrow 0 \text{ as } n, m \rightarrow \infty.$$

$\Rightarrow \{z_n\}$ is a Cauchy seq. By closeness $V \subset H$.

Rmk: For $1 < p < \infty$, the "existence of minimizer" still holds in $L^p(X)$,
 but fails for $L^1(X)$, $L^\infty(X)$.

Cor: Let H be a Hilbert Space. $V \cap H$ is a closed vector space of H ; then $\forall x \in H$. \exists unique $x_V \in V$ and $x_{V^\perp} \in V^\perp = \{w \in H : \langle v, w \rangle = 0, \forall v \in V\}$.
 s.t: $x = x_V + x_{V^\perp}$

Moreover x_V is the vector s.t: $\|x - x_V\| = \inf \{ \|x - z\|, z \in V\}$

Proof: Only need to prove $x_{V^\perp} : x - x_V \in V^\perp$

Suppose $x_{V^\perp} \notin V^\perp$, i.e. $\exists v \in V$ s.t. $\langle x_{V^\perp}, v \rangle \neq 0$.

$\Rightarrow \exists v \in V$ s.t. $\langle x_{V^\perp}, v \rangle = 1$.

$$\begin{aligned} \text{Now. } \|x - (x_V + \varepsilon v)\|^2 &= \|x_{V^\perp} - \varepsilon v\|^2 = \|x_{V^\perp}\|^2 + \varepsilon^2 \|v\|^2 - 2\varepsilon \\ &= \|x - x_V\|^2 - 2\varepsilon + \varepsilon^2 \|v\|^2 \\ &< 0 \quad \varepsilon \rightarrow 0 \end{aligned}$$

$$\Rightarrow \|x - (\pi v + \epsilon v)\| < \|x - \pi v\| \text{ contradiction.}$$

$H = V \oplus V^\perp$

Recall: \mathbb{R}^d .

$$\begin{aligned} L: \mathbb{R}^d &\rightarrow \mathbb{R} \text{ is linear functional} \\ \Leftrightarrow L_x &= a_1 x_1 + \dots + a_d x_d. \\ \downarrow a &= (a_1, \dots, a_d) \in \mathbb{R}^d. \end{aligned}$$

Now let H be a Hilbert space. For any $v \in H$, we define a linear functional $L_v: H \rightarrow \mathbb{R}$

Note: L_v is continuous

$$|L_v(w - w_i)| = |\langle v, w - w_i \rangle| \leq \|v\| \cdot \|w - w_i\|.$$

Thm (Riesz representation thm)

Let H be a Hilbert space

$L: H \rightarrow \mathbb{R}$ is any continuous linear functional

Then there exists a unique $v \in H$

$$\text{s.t. } L_v = L$$

Space of all continuous linear functional on $H \cong H$.

Proof: (Uniqueness)

$$\text{If } L_{v_1} = L_{v_2} = L$$

$$\Rightarrow L_{v_1}(v_1 - v_2) = L_{v_2}(v_1 - v_2)$$

$$\Rightarrow \langle v_1, v_1 - v_2 \rangle = \langle v_2, v_1 - v_2 \rangle$$

$$\Rightarrow \|v_1 - v_2\| = 0.$$

(Existence) WLOG, we assume $L \neq 0$.

Let $V = \ker L = \{x \in H : L(x) = 0\} = L^{-1}(0)$ closed.

$$\Rightarrow \underline{V^\perp \neq 0}.$$

Take $w \in V^\perp$ s.t. $\|w\| = 1 \Rightarrow \langle w \rangle \neq 0$.

Let $v = \langle w \rangle w$. Then.

$$\begin{aligned} L_v(x) &= \langle \langle w \rangle w, x \rangle = \langle \langle w \rangle w, \underbrace{x - \frac{\langle x \rangle}{\langle w \rangle} w}_{V} + \underbrace{\frac{\langle x \rangle}{\langle w \rangle} w}_{V^\perp} \rangle = \langle \langle w \rangle w, \frac{\langle x \rangle}{\langle w \rangle} w \rangle \\ &= \langle x \rangle. \quad \forall x \in H. \end{aligned}$$

Lee 23

Today: linear functionals on $L^1(X)$ (function defined on abstract space
function space)

1. Linear functionals on normed vector spaces.

We have seen.

Riesz representation thm for $L^2(X)$: $L: L^2(X) \rightarrow \mathbb{R}$.

continuous linear functional

$\Rightarrow \exists! g \in L^2(X)$ s.t. $L(f) = \int_X f \cdot g \, d\mu. \quad \forall f \in L^2(X)$.

Note: If we use complex-valued functions i.e.

$L^2(X, \mathbb{C})$, then $\exists! g \in L^2(X, \mathbb{C})$ s.t. $L(f) = \int_X f \bar{g} \, d\mu$.

Riesz representation thm for $C_c(X)$

$L: C_c(X) \rightarrow \mathbb{R}$ be bounded linear functional

$\Rightarrow \exists$ Borel measure μ_1, μ_2 on X s.t.: $L(f) = \int_X f \, d\mu_1 - \int_X f \, d\mu_2 = \int_X f \, (d\mu_1 - d\mu_2)$ signed measure.

Note: We proved this for positive linear functionals in $L^1(\mu_1, \mu_2 = 0)$ and then in PSet 10-1 for general L (with X compact).

Note: assumption on X is: X is locally compact metric space, and it is σ -compact

↑
only need:
Hausdorff

• Justification of assumptions/definitions.

(1) The spaces $L^2(X)$, $C_c(X)$ and $L^p(X)$ are normed vector spaces:

$$L^2(X): \|f\|_2 = \left(\int_X |f|^2 d\mu \right)^{1/2}$$

$$C_c(X): \|f\|_{C_c(X)} = \sup_{x \in X} |f(x)|$$

$$L^p(X): \|f\|_{L^p(X)} = \left(\int_X |f|^p d\mu \right)^{1/p}.$$

(2) Let V be a normed vector space: $\ell: V \rightarrow \mathbb{R}$ linear functional

Then ℓ is continuous $\iff \ell$ is continuous at 0 $\iff \ell$ is bounded

\forall open set $U \subset \mathbb{R}$

$\ell^{-1}(U)$ is open

(a)

$\exists r > 0$, $\ell^{-1}((-r, r))$ contains

an open set (centered on 0),

(b)

$\exists c \text{ s.t.}$

(c)

Reason: (a) \Rightarrow (b), (c) \Rightarrow (b)

We also have (b) \Rightarrow (a): Let $v \in \ell^{-1}(U) \Rightarrow \exists \varepsilon \text{ s.t.}$

$(\ell(v)-\varepsilon, \ell(v)+\varepsilon) \subset U \Rightarrow$ By triangle inequality $v \in \ell^{-1}((-r, r))$ $\underset{\text{open } \ell^{-1}(U)}{\text{open}}$

Finally (a) \Rightarrow (c) By (a) $\ell^{-1}((-1, 1))$ is open.

Since $0 \in \ell^{-1}((-1, 1))$, so $\exists r > 0$ s.t. $B_r(0) \subset \ell^{-1}((-1, 1))$.

i.e. $\ell(B_r(0)) \subset (-1, 1) \Rightarrow |\ell(r)| = \frac{\|r\|}{r} |\ell(\frac{r}{\|r\|})| \leq \frac{\|r\|}{r}$

$$f \geq 0 \Rightarrow \ell(f) \geq 0.$$

(3) Any positive linear functional, $\ell: C(X) \rightarrow \mathbb{R}$ is bounded if X is compact.

Reason: Let $c = |\ell(1)|$. for $\forall f \in C(X)$ and $|f| \leq 1$.

We have $1 \pm f \geq 0 \Rightarrow c = \ell(f) \geq 0 \Rightarrow |\ell(f)| \leq c$.

So for any $f \in C(X)$,

$$|\ell(f)| = \left| \ell \left(\frac{f}{\|f\|} \cdot \|f\| \right) \right| = \|f\| \cdot \left| \ell \left(\frac{f}{\|f\|} \right) \right| \leq \|f\| \cdot c.$$

Note: By the same argument one can prove that if X is locally compact metric (or Hausdorff) space then for any compact set $K \subset X \exists c_K \text{ s.t.}$

$$|\ell(f)| \leq c_K \|f\|. \quad \forall f \in C_c(X) \text{ with } \text{supp}(f) \subset K.$$

for positive linear functional ℓ .

2. Linear functionals on $L^p(X)$

Now consider the space $L^p(X)$, $p \geq 1$.

As in the case of $L^2(X)$, one can start with a function g and define $(Lg): L^p(X) \rightarrow \mathbb{R}$ $f \mapsto \int_X f \cdot g \, d\mu$.

Note that for Lg to be well-defined, we need $f \cdot g \in L^1(X)$ for all $f \in L^p(X)$. This is the case if $g \in L^{\frac{1}{p}}(X)$ where $\frac{1}{p} + \frac{1}{q} = 1$.

Prop: // for any $g \in L^{\frac{1}{p}}(X)$, $Lg: L^p(X) \rightarrow \mathbb{R}$ is a continuous linear functional.

Proof: We have seen that Lg is well-defined.

It's obviously linear i.e. $Lg(c_1f_1 + c_2f_2) = c_1Lg(f_1) + c_2Lg(f_2)$.

To prove Lg is continuous, it's enough to prove

Lg is bounded, which follows from the Hölder's inequality,

$$|Lg(f)| \leq \|fg\|_{L^1} \leq \|f\|_{L^p} \|g\|_{L^{\frac{1}{p}}} = C \|f\|_{L^p}. \quad C = \|g\|_{L^{\frac{1}{p}}}.$$

vector space the main theorem we want to prove today is

~~continuous~~ ~~functional~~ Riesz representation theorem for $L^p(X)$

~~all linear~~ Suppose μ is a σ -finite measure on X .

~~functional on~~ (i.e. $X = \bigcup_{n=1}^{\infty} X_n$ and $\mu(X_n) < \infty$, H_n).

$(L^p)^* = L^q$. Let $1 \leq p < \infty$ and $q \geq 1$ s.t. $\frac{1}{p} + \frac{1}{q} = 1$.

$\|Lg\|_{L^q} = \|g\|_{L^{\frac{1}{p}}}$. Then for any continuous linear functional $\ell: L^p(X) \rightarrow \mathbb{R}$, there exists a unique $g \in L^{\frac{1}{p}}(X)$ s.t. $\ell = Lg$.

Rank: • The theorem fails for $p = +\infty$.

In other words, for any $g \in L^1(X)$, g defines a continuous linear functional $Lg: L^\infty(X) \rightarrow \mathbb{R}$, however, there are more continuous linear functionals on $L^\infty(X)$ than these Lg 's.

• The main ingredient in the proof is the following.

theorem that we will prove later

Radon-Nikodym Thm:

- || Let μ be a σ -finite measure on (X, \mathcal{F}) .
- || Let ν be a signed measure on $(X, \mathcal{F}) \rightarrow$ i.e. $\nu: \mathcal{F} \rightarrow \mathbb{R}$
s.t. $\nu(\emptyset) = 0$.
- || s.t. $\mu(A) = 0 \Rightarrow \nu(A) = 0$.
- || Then $\exists f \in L^1(X)$ s.t. $\nu(A) = \int_A f d\mu$
- || $L^1(X) = \{f \in L^1(X) \mid \int_X f d\mu = 0\}$

$$\nu(A_n) = \sum_{A \in \mathcal{F}, A \subset A_n} \nu(A)$$

disjoint

• Proof of Riesz representation thm for $L^p(X)$

[Step 1] (Uniqueness) Let $g_1, g_2 \in L^2(X)$ s.t. $Lg_1 = Lg_2$

$$\Rightarrow Lg_1 - g_2 = 0 \text{ on } L^2(X).$$

Let $g = g_1 - g_2 \in L^2(X)$. Suppose $g \neq 0$ in $L^2(X)$. then \exists measurable set $A \subset X$

with $0 < \mu(A) < \infty$ s.t. $g \geq 0$ on A or $g \leq 0$ on A .

Let $f = \chi_A \in L^p(X)$, Then $Lg(f) = \int_X f \cdot g d\mu \neq 0$. Contradiction.

[Step 2] Suppose $\mu(A) < \infty$.

[Idea: linear functional \rightarrow signed measure $\xrightarrow{R-N}$ a function $g \in L^1$.
 $g \rightarrow L^2$.]

Since $\mu(X) < \infty$ for any measurable set $A \in \mathcal{F}$. Set $A \in \mathcal{F}$.
the function $\chi_A \in L^p(X)$. So we can define $\nu: \mathcal{F} \rightarrow \mathbb{R}$ by

$$\nu(A) := \ell(\chi_A)$$

This is a signed measure since

$$\textcircled{1} \quad \nu(\emptyset) = \ell(0) = 0$$

\textcircled{2} Let $A_1, A_2, \dots \in \mathcal{F}$ be disjoint. Then $\chi_{\bigcup_{i=1}^n A_i} \rightarrow \chi_{\bigcup_{i=1}^\infty A_i}$ pointwise. Also $|\chi_{\bigcup_{i=1}^K A_i}| \leq 1 \in L^p(X)$.

So by dominated convergence theorem $\chi_{\bigcup_{i=1}^K A_i} \rightarrow \chi_{\bigcup_{i=1}^\infty A_i}$ in $L^p(X)$.
Since ℓ is continuous, we get

$$\sum_{i=1}^k \ell(\chi_{A_i}) = \ell(\chi_{\bigcup_{i=1}^k A_i}) \rightarrow \ell(\chi_{\bigcup_{i=1}^\infty A_i}).$$

$$\text{i.e. } \nu(\bigcup_{i=1}^\infty A_i) = \lim_{k \rightarrow \infty} \sum_{i=1}^k \nu(A_i).$$

• By Radon-Nikodym theorem, $\exists g \in L^1(X)$ s.t.

$$\nu(A) = \int_A g d\mu \quad \forall A \in \mathcal{F} \quad \text{Note: } \mu(A)=0 \Rightarrow X_A=0 \text{ in } L^p(X) \Rightarrow \nu(A) = \ell(X_A)=0$$

We want to prove $\ell = Lg$ on $L^p(X)$.

By def. for any $A \in \mathcal{F}$, one has $\ell(X_A) = \nu(A) = \int_A g d\mu = Lg(X_A)$.

By linearity, $\ell = Lg$ for all simple function.

In Lec. 21, we have shown that the space of simple functions is dense in $L^p(X)$.

Reason: Assume $g \geq 0$, otherwise consider g_+ and g_-

Let $f \in L^p(X)$. Assume $f \geq 0$. Otherwise consider f_+ and f_-

Then \exists simple functions $\varphi_1, \varphi_2, \dots \xrightarrow{\text{f in } L^p(X)}$

Assume $\|f\| \leq c$. Then f By dominated convergence,

$$Lg(f) = \lim_{n \rightarrow \infty} Lg(\varphi_n) = \lim_{n \rightarrow \infty} \ell(\varphi_n) = \ell(f) < \infty \Rightarrow Lg \text{ well defined}$$

For unbounded $f \in L^p(X)$, $f \geq 0$, we let $f_N = \min(f, N)$.

Then by monotone convergence, $Lg(f) = \lim_{N \rightarrow \infty} Lg(f_N)$

$$= \lim_{N \rightarrow \infty} \ell(f_N) = \ell(f) < \infty.$$

It remains to prove $g \in L^q$. WLOG, we assume $g \geq 0$.

Since $L = Lg$ is continuous, it's a bounded linear functional.

So $\exists c > 0$ s.t: $|Lg(f)| \leq c \|f\|_{L^p(X)} \quad \forall f \in L^p(X)$.

Case 1: $1 < p < \infty$.

Idea: Want $\|g\|_{L^q} < \infty$. So we take $f = g^{q-1}$

so that formally $\frac{\|g\|_{L^q}}{q} = |Lg(f)| \leq c \|f\|_{L^p}$

$$\text{and } \|f\|_{L^p} = \left(\int_X |g|^{q-1} d\mu \right)^{\frac{1}{q-1}}$$

$$= \left(\int_X |g|^{q-1} d\mu \right)^{\frac{1}{q-1}} = \frac{\|g\|_{L^q}}{q}$$

$\Rightarrow \|g\|_{L^q} \leq c$. But this argument is NOT true

since we don't know $f = g^{q-1} \in L^p$ yet

Let $f_N = \min(g, N)^{q-1}$ Then $f_N \in L^p(X)$.

Since $\mu(X) < \infty$, we have $\underline{Lg}(f_N) = \int_X g f_N d\mu$.

$$\geq \int_X \min(g, N)^{q-1} d\mu = \underline{\|\min(g, N)\|_{L^q}}.$$

$$\Rightarrow \|f_N\|_{L^p} = \left(\int_X \lim_{N \rightarrow \infty} (g_N)^p d\mu \right)^{1/p} = \left\| \lim_{N \rightarrow \infty} g_N \right\|_{L^p}^{p-1}$$

$$\Rightarrow \left\| \lim_{N \rightarrow \infty} g_N \right\|_{L^p} \leq c.$$

Letting $N \rightarrow \infty$. Using monotone convergence, we get
 $\|g\|_{L^p} \leq c.$

Case 2: $p=1$

Let $f_N = \chi_{g > N}$. Then $f_N \in L^1(X)$. So.

$$\begin{aligned} N \int g > N 1 d\mu &\leq \int g > N g d\mu = \int g f_N d\mu \\ &\leq c \|f_N\|_{L^1} = c \int g > N 1 d\mu. \end{aligned}$$

\Rightarrow For $N > c$. $\mu(\{g > N\}) = 0$. So $\|g\|_{L^\infty} \leq c$.

Step 3 Suppose $\mu(X) = +\infty$

By assumption $\exists x_1 < x_2 \subset X$. $\mu(X_n) < \infty$.

$$\text{s.t. } X = \bigcup_{n=1}^{\infty} X_n.$$

By Step 2. one can find $g_n \in L^2(X_n)$ s.t. $g_n \perp g_m$ on $L^p(X_n)$.

By Step 1. for each n one has $g_m = g_n$ on X_m . Thus
 one get one function g defined on X s.t.

$$g = g_n \text{ on } X_n$$

In Step 2. we showed that $\|g_n\|_{L^2(X_n)} \leq c$. for some c, b_n
 So by monotone convergence argument
 $g \in L^2(X)$ and $\|g\|_{L^2} \leq c$.

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1. Signed measure

- Recall (Lee 10.11.16)

Countable additivity $\int_{\bigcup A_n} f d\mu = \sum_{n=1}^{\infty} \int_{A_n} f d\mu$ A_n 's disjoint

• If f is non-negative, then " $\mu_f = f d\mu$ " is a measurable, where $\mu_f(A) = \int_A f d\mu$

• if f is absolutely integrable, then " $\mu_f = f d\mu$ " is a "signed measure", where $\mu_f(A) = \int_A f d\mu$.

Rmk. • Measure could take value $+\infty$

• Signed measure could take value $+\infty$ or $-\infty$.

So we don't really need to assume $f \in L^1$ to construct μ_f

A problem. If $A_1, A_2 \subset X$, disjoint and $\mu(A_1) = -\infty$, $\mu(A_2) = +\infty$

Q: What is $\mu(A_1 \cup A_2)$?

Need to assume μ can not attain both $+\infty$ and $-\infty$.

Def: || A signed measure on a measurable space (X, \mathcal{F})
is a function $\mu: \mathcal{F} \rightarrow [-\infty, +\infty]$.

S.t: (1) $\mu(\emptyset) = 0$.

(2) $\mu(\mathcal{F}) \subset [-\infty, +\infty]$ or $(-\infty, +\infty]$

(3) $\mu\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} \mu(A_n)$ for disjoint $A_n \in \mathcal{F}$.

Moreover, if $\mu\left(\bigcup_{n=1}^{\infty} A_n\right) \neq +\infty$. We require the series $\sum_{n=1}^{\infty} \mu(A_n)$ converges absolutely

Example: (1) If μ_1, μ_2 are measures, either μ_1, μ_2 is finite. ^{or}
Then $\mu = \mu_1 - \mu_2$ is a signed measure.

(2) Let f be a measurable function on X .

Suppose either $f_+ \in L^1(X, \mu)$ or $f_- \in L^1(X, \mu)$.

then " $\mu_f = f d\mu$ " is a signed measure

$$\mu_f(A) = \int_A f d\mu = \int_A f_+ d\mu - \int_A f_- d\mu.$$

Integrable in the extended sense.

Def: Let μ be a signed measure on (X, \mathcal{F})

(1) We say $A \in \mathcal{F}$ is a positive set if \forall measurable $B \subset A$, $\mu(B) \geq 0$.

(2) \dots negative set $\dots \mu(B) \leq 0$

(3) \dots null set $\dots \mu(B) = 0$

Hahn decomposition theorem | Let μ be a signed measure on (X, \mathcal{F}) . Then \exists positive set X_+ and negative set X_- for μ .
s.t.: $X = X_+ \vee X_-$, $X_+ \cap X_- = \emptyset$. Moreover if X'_+, X'_- is another such decomposition, then $X'_+ \triangle X'_-$ is a null set
 $X'_- \triangle X'_+$

Proof: WLOG assume $\mu(A) < +\infty$.

(Idea: Pick X_+ to be a positive set with maximal measure)
Let $m_+ = \sup \{\mu(A) \mid A \text{ is a positive set}\} < +\infty$.

Let A_1, A_2, \dots be a sequence of positive sets s.t.
 $\mu(A_n) \rightarrow m_+$.

Let $A = \bigcup_{n=1}^{\infty} A_n$ Then A is a positive set and

$$\mu(A) = \lim_{n \rightarrow \infty} \mu(A_n) = m_+$$

Let $X_+ = A$, $X_- = A^c$ Need: X_- is a negative set.

By contradiction: Suppose $\exists B_1 \subset X_-$, s.t. $\mu(B_1) > 0$.

If B_1 is a positive set, then $\mu(X_+ \cup B_1) > \mu(X_+)$
if $A_1 \subset A_2 \subset \dots$ then $\mu(\bigcup A_n) = \lim_{n \rightarrow \infty} \mu(A_n)$.

If B_1 is NOT a positive set. Then $\exists B_2 \subset B_1$. s.t:

$$\mu(B_2) > \mu(B_1). \Rightarrow \exists n, \text{ s.t. } \mu(B_2) > \mu(B_1) + \frac{1}{n}.$$

Greedy Choice Take B_2, n , s.t: n is as small as possible

Again: If B_2 is positive, then we have done.

→ We get a sequence $B_2, n_1; B_3, n_2; B_4, n_3; \dots$ s.t:

$$B_1 \supset B_2 \supset B_3 \supset \dots \quad \mu(B_{k+1}) > \mu(B_k) + \frac{1}{nk} \quad nk \text{ is the smallest poss.}$$

$$\text{Let } B = \bigcap_{i=1}^{\infty} B_i \quad \text{Then } +\infty > \mu(B) = \lim_{n \rightarrow \infty} \mu(B_n) = \sum_{i=1}^{\infty} \frac{1}{ni}$$

$$\text{So } \lim_{i \rightarrow \infty} n_i = +\infty$$

Claim: B is a positive set

If NOT: $\exists C \subset B$ s.t: $\mu(C) > \mu(B)$.

$$\Rightarrow \exists m_0 > 0 \text{ and } N_0 > 0 \text{ s.t. } \forall k > N_0. \mu(C) - \mu(B_k) > \frac{1}{m_0}$$

Take k s.t: $n_k > m_0$.

$\mu(C) > \mu(B_k) + \frac{1}{m_0}$ contradiction with the choice of B_{k+1} !

$$\text{Suppose } X = X_+ \vee X_- = X'_+ \vee X'_-$$

$$\text{Then } \underbrace{X_+ \Delta X'_+}_{\text{positive}} = \underbrace{X_- \Delta X'_-}_{\text{negative.}} \rightarrow \text{Null.}$$

e) Then (Jordan decomposition theorem)

Let μ be a signed measure on (X, \mathcal{F}) . then \exists unique measures μ_+, μ_- s.t: $\mu = \mu_+ - \mu_-$ i.e. $\mu_+ \mid_{X_-} = 0$ and $\mu_- \mid_{X_+} = 0$.

Moreover if $X = X_+ \vee X_-$ is a Hahn decomposition, then

$$\mu_+ = \mu \mid_{X_+}, \quad \mu_- = -\mu \mid_{X_-}$$

$$\text{i.e. } \forall A \in \mathcal{F}, \mu_+(A) = \mu(A \cap X_+)$$

$$\mu_-(A) = -\mu(A \cap X_-)$$

Proof. $\mu_+(A), \mu_-(A)$.

One can check. μ_+, μ_- are measure and $\mu = \mu_+ - \mu_-$
one is finite.

Suppose we also have $\mu = \nu_+ - \nu_-$ and $\nu_+|_{X_-} = 0$, $\nu_-|_{X_+} = 0$.
 Then $A \subset X_+$, we have $\mu(A) = \nu_+(A)$.
 $\Rightarrow \forall A \in \mathcal{F}$.

$$\mu_+(A) = \mu(A \cap X_+) = \mu(A \cap X'_+) = \nu_+(A \cap X'_+) = \nu_+(A).$$

$$\text{Similarly, } \mu_-(A) = \nu_-(A).$$

Rank. • In the Jordan decomposition, $\mu = \mu_+ - \mu_-$.

We call $\mu_+(\mu_-)$ the positive (negative) part of μ .

or positive (negative) variation of μ .

def 1. We can define the absolute value (or total variation) of μ to be $|\mu| = \mu_+ + \mu_-$

* def 2. We call a signed measure μ is finite or σ -finite if $|\mu|$ is finite or σ -finite.

def 3. In general, we say two signed measure μ, ν are mutually
singular
 $\exists X = X_1 \sqcup X_2$, s.t: $\mu|_{X_2} = 0$, $\nu|_{X_1} = 0$. Notation: $\mu \perp \nu$.
 $\Leftrightarrow \mu \text{ supported on } X_1$, $\nu \text{ supported on } X_2$

def 4. We say ν is absolutely continuous w.r.t μ .
 if every μ -null set is also a ν -null set.
 Notation: $\nu \ll \mu$.

Example: • Jordan decomposition $\mu = \mu_+ - \mu_-$, $\mu_+ \perp \mu_-$
 • Dirac measure $\delta_{x_0} \perp$ lebesgue measure m
 • Let f be a measurable function which is integrable in extended sense.

$\rightarrow \mu_f = f d\mu$ is a signed measure and $\mu_f \ll \mu$

Prop: Let ν be a finite signed measure, and let μ be a (positive) measure on (X, \mathcal{F}) . Then $\nu \ll \mu$ if and only if $\forall \varepsilon > 0, \exists \delta > 0$ s.t: $\forall A \in \mathcal{F}$, if $\mu(A) < \delta$ then $|\nu(A)| < \varepsilon$.

Lebesgue-Radon-Nikodym Theorem

Let μ be a σ -finite measure, and ν a σ -finite signed measure. Then there exists unique signed measures μ_r, μ_s s.t. $\nu = \mu_r + \mu_s$

where $\mu_s \perp \mu$. $\mu_r << \mu$

Moreover \exists (extended sense) integrable function f on X s.t. $d\mu_r = f d\mu$.

Rmk: if ν is measure, then so are μ_r, μ_s .
if ν is finite, then μ_r is finite.
 μ_r is a measure $\Leftrightarrow f \geq 0$.

Cor. (Radon-Nikodym Theorem)

Suppose $\nu << \mu$. ($\mu(A) = 0 \Rightarrow \nu(A) = 0$).
Then \exists integrable function f s.t. $d\nu = f d\mu$.
Moreover, if ν is finite, then $f \in L^1(X)$

Proof: if $\nu \leq \mu$, $\nu \perp \mu$ then $\nu = 0$.

we call f the Radon-Nikodym derivative of ν w.r.t. μ .

$$d\nu = f d\mu \quad f = \frac{d\nu}{d\mu}$$

Proof: Case 1 ν is finite measure

Define $M = \{f: X \rightarrow [0, +\infty] : \int_A f d\mu \leq \nu(A), \forall A \in \mathcal{F}\}$.

Then $M \neq \emptyset$ since $0 \in M$

If $f, g \in M$, then $\max(f, g) \in M$.

Reason: Let $B = \{f > g\}$. Then $\complement A \in \mathcal{F}$.

$$\int_A \max(f, g) d\mu = \int_{A \cap B} f d\mu + \int_{A \cap B^c} g d\mu \leq \nu(A \cap B) + \nu(A \cap B^c) \leq \nu(A)$$

Now set $m := \sup_{f \in M} \int_A f d\mu \leq \nu(A) < +\infty$

Let $f_n \in M$ s.t. $\int_A f_n d\mu \rightarrow m$.

Let $g_n = \sup(f_1, \dots, f_n) \in M$. Then $g_n \nearrow f := \sup_n f_n$
 $\Rightarrow m = \lim_{n \rightarrow \infty} \int_A f_n d\mu \leq \lim_{n \rightarrow \infty} \int_A g_n d\mu \leq m$.

by monotone convergence $\int_A f d\mu = m$.
 $\Rightarrow f$ is a.e. finite

WLOG, assume $f \geq 0$

Note: $m = \int_A f d\mu \leq V(A)$ i.e. $f \in M$.

Let $\mu_s = \nu - \frac{f}{\mu} d\mu$.

Need: μ_s is singular w.r.t. μ .

Lemma: || Let μ, ν be finite measures on (X, \mathcal{F})

Then either $\mu \perp \nu$, or $\exists \varepsilon > 0$ and $A \in \mathcal{F}$,

s.t: $\nu(A) > 0$ and A is a positive set w.r.t. $\mu \ll \nu$.

Proof: By lemma $\exists \varepsilon > 0$, $A \in \mathcal{F}$ s.t: $\mu(A) >$

$$\forall B \subset A \quad \mu_s(B) > \varepsilon \mu(B)$$

In particular, $\forall C \in \mathcal{F}$.

$$\int_C (f + \varepsilon \chi_A) d\mu = \int_C f d\mu + \varepsilon \mu(A \cap C)$$

$$\leq \int_C f d\mu + \nu(A \cap C) - \int_{A \cap C} f d\mu.$$

$$= \int_{C \cap A^c} f d\mu + \nu(A \cap C)$$

$$\leq \nu(C \cap A^c) + \nu(A \cap C) = \nu(C).$$

i.e. $f + \varepsilon \chi_A \in \mu$.

i.e. $m \geq \int (f + \varepsilon \chi_A) d\mu = \int f d\mu + \varepsilon \int \chi_A d\mu = m + \varepsilon \mu(A) > m$.
 contradiction.

Thus $\mu_s \perp \mu$.

Let $\mu_r = f d\mu$. Then $\nu = \mu_r + \mu_s$ $\mu_r \ll \mu$, $\mu_s \perp \mu$.
 Moreover $f \geq 0$ and $f \in L^1(X, \mu)$

Uniqueness $\nu = \mu_r' + \mu_s' = \mu_r + \mu_s$.

$$\Rightarrow \mu_r' - \mu_r = \mu_s - \mu_s'$$

$$\mu_r' - \mu_r \ll \mu. \quad \mu_s - \mu_s' \perp \mu$$

$$\Rightarrow \mu_r' - \mu_r = \mu_s - \mu_s' = 0.$$

Lec 25 Fundamental theorem of Calculus (Version 1).

Case of dimension 1

Recall. $F: (a,b) \rightarrow \mathbb{R}$. \rightarrow We call a function f the derivative of F if: $f(x) = \lim_{h \rightarrow 0} \frac{F(x+h) - F(x)}{h}$

We learned in Calculus

If f is a continuous function, we can define

$$F(x) = \int_0^x f(t) dt. \text{ Then } F'(x) = f(x).$$

In other words, we have

$$f(x) = \lim_{h \rightarrow 0^+} \frac{1}{h} \int_x^{x+h} f(t) dt = \lim_{h \rightarrow 0^+} \frac{1}{h} \int_{x-h}^x f(t) dt$$

Thm. (Lebesgue differentiation theorem dim=1 case).

Take $\mu = m$.

Suppose $f \in L^1(\mathbb{R})$ then

$$\lim_{h \rightarrow 0^+} \frac{1}{h} \int_{[x, x+h]} f(t) dt = f(x) \text{ a.e. } x \in \mathbb{R}$$

$$\lim_{h \rightarrow 0^+} \frac{1}{h} \int_{[x-h, x]} f(t) dt = f(x) \text{ a.e. } x \in \mathbb{R}$$

i.e. If we let $F(x) = \int_{(-\infty, x]} f(t) dt$, then $F'(x) = f(x)$. a.e. $x \in \mathbb{R}$

Idea.

Density argument

We want to prove some property say. $\lim_{n \rightarrow \infty} L_n(f) = g$.

for a class of functions f

Step 1. Prove that the formula holds for a dense sub-class of nice functions

Step 2. Control the difference $|L_n(f_1) - L_n(f_2)| \leq \text{some norm of } f-g$

Step 3. Combine 2 steps to get the full proof

for \mathbb{R}^d

Thm ((One-side) Hardy-Littlewood maximal inequality)

Suppose $f \in L^1(\mathbb{R}^d)$ then for $\delta > 0$.

$$m(\{x \in \mathbb{R}^d : \sup_{h>0} \frac{1}{h} \int_{[x, x+h]} |f(t)| dt > \alpha\}) \leq \frac{1}{\alpha} \int_{\mathbb{R}^d} |f(t)| dt.$$

(higher dimension)

$$m(\{x \in \mathbb{R}^d : \frac{\sup_{r>0} \frac{1}{m(B(x,r))} \int_{B(x,r)} |f(t)| dt}{\maximal \text{ fraction}} > \alpha\}) \leq \frac{3^d}{\alpha} \int_{\mathbb{R}^d} |f(t)| dt.$$

$f^*(x)$

(*)

~~$\frac{1}{h} \int_{[x, x+h]} f(x) dt - f(x) = \frac{1}{h} \int_{[x, x+h]} f(t) - f(x) dt$~~

Proof. We only need to prove (1). ~~$|f(t) - f(x)| \leq \frac{|f(t) - g(t)| + |g(t) - f(x)|}{2}$~~

We have known that (1) holds for continuous f .

[PSet 6-1 problem 2(3)] $\forall f \in C^1(K)$ $\forall \varepsilon > 0 \exists g$ continuous with compact support s.t. $\|f - g\|_{L^1} = \int_K |f(x) - g(x)| dm \leq \varepsilon$.

By (*) $m(\{x: \sup_{h>0} \frac{1}{h} \int_{[x, x+h]} |f(t) - g(t)| dt > \alpha\}) \leq \frac{\varepsilon}{2}$.
By Markov:

$$m(\{x: |f - g| > \alpha\}) \leq \frac{\varepsilon}{2}.$$

So $\exists E$ with $m(E) \leq \frac{2\varepsilon}{\alpha}$ s.t. $\forall x \in R \setminus E$.

$$\frac{1}{h} \int_{[x, x+h]} |f(t) - g(t)| dt \leq \alpha.$$

$$|f(x) - g(x)| \leq \alpha \rightarrow \frac{1}{h} \int_{[x, x+h]} |f(x) - g(x)| dt \leq \alpha$$

By calculus $\frac{1}{h} \int_{[x, x+h]} (g(t) - g(x)) dt \leq \alpha$.

$$\left| \frac{1}{h} \int_{[x, x+h]} g(t) dt - g(x) \right| \leq \alpha \text{ for all } h \text{ small enough.}$$

By triangle inequality, for all h small enough.

$$\left| \frac{1}{h} \int_{[x, x+h]} f(t) dt - f(x) \right| \leq 3\alpha.$$

$$\Rightarrow \limsup_{h \rightarrow 0} \left| \frac{1}{h} \int_{[x, x+h]} f(t) dt - f(x) \right| \leq 3\alpha.$$

keeping α fixed. let $\varepsilon \rightarrow 0$.

$$\limsup_{h \rightarrow 0} \left| \frac{1}{h} \int_{[x, x+h]} f(t) dt - f(x) \right| \leq 3\alpha \quad \text{A.e. } m(A_\alpha^c) = 0 \quad \text{A.e. } x \in R \text{ a.e.}$$

$$\text{Let } \alpha = \frac{1}{h} \rightarrow 0.$$

$$m(\bigcup_n A_{\frac{1}{n}}^c) = 0$$

$$\frac{1}{h} \int_{[x, x+h]} f(t) dt = f(x) \text{ a.e. } x.$$

$R \rightarrow R^d$. Suppose $f \in C^1_{loc}(R^d)$

$$(1) \lim_{r \rightarrow 0} \frac{1}{m(B_r(x))} \int_{B_r(x)} f(t) dt = f(x) \quad \text{a.e. } x \in R^d$$

$$\lim_{h \rightarrow 0} \frac{1}{2h} \int_{[x-h, x+h]} f(t) dt = f(x)$$

Goal:
 ii) Moreover, $\lim_{r \rightarrow 0} \frac{1}{m(B(x,r))} \int_{B(x,r)} |f(x) - f(t)| dt = 0 \text{ a.e. } x \in \mathbb{R}^d$

Note: In dim 1. $\lim_{h \rightarrow 0} \frac{1}{2h} \int_{[x-h, x+h]} f(t) dt = F'(t)$.

So one can think of

" $\lim_{r \rightarrow 0} \frac{1}{m(B(x,r))} \int_{B(x,r)} f(t) dt$ " as a "derivative"
 $\frac{m(B(x,r))}{m(B(x,r))}$ Let $d\mu = f dm$

"derivative" of μ w.r.t. m .

Remark: The theorem holds for $f \in L^1_{loc}(\mathbb{R}^d)$.

i.e. $\forall x \exists r \text{ s.t. } f \in L^1(B(x,r))$

Notation: $Mf := \sup_{r > 0} \frac{1}{m(B(x,r))} \int_{B(x,r)} |f(t)| dt$

"Vitali-type Covering Lemma": Let $\mathcal{U} = \{B(x_\alpha, r_\alpha)\}$ be a family of open balls, s.t: $\bigcup \mathcal{U} \subset \bigcup B(x_\alpha, r_\alpha)$. Then $\forall c < m(\mathcal{U})$

one can find disjoint balls $B(x_1, r_1) - B(x_n, r_n)$ in \mathcal{U} s.t:

$$\sum_{j=1}^n m(B(x_j, r_j)) > 3^{-d} c.$$

Proof of Hardy-Littlewood

Let $U = \{x \in \mathbb{R}^d : Mf > \alpha\}$.

For any $x \in U$, $\exists r(x)$ s.t: $\int_{B(x, r(x))} |f(t)| dt > \alpha \cdot m(B(x, r(x)))$

Then $U \subset \bigcup B(x_i, r_i)$

by Vitali, $\forall c < m(U)$, $\exists B(x_1, r_1), \dots, B(x_n, r_n)$ in this family
 s.t: $c < 3^d \sum_{j=1}^n m(B(x_j, r(x_j))) < 3^d \frac{1}{2} \sum_j \int_{B(x_j, r(x_j))} |f(t)| dt$
 $\leq \frac{3^d}{2} \int_{\mathbb{R}^d} |f(t)| dt$ disjoint.

Proof of Vitali properties in
↓ Radon measure

For any $c < m(U)$ one can find K compact s.t:

$m(K) > c$. $K \subset U$. Since $K \subset \bigcup_{\alpha} B(x_{\alpha}, r_{\alpha})$, one can find $B(y_1, s_1), \dots, B(y_m, s_m)$ in the family \mathcal{U} s.t:

$$K \subset \bigcup_{i=1}^m B(y_i, s_i)$$

WLOG assume $s_1 \geq s_2 \geq \dots \geq s_m$.

"Greedy algorithm": Let $A_1 = B(y_1, s_1)$.

Suppose we have got A_1, \dots, A_{j-1} , let A_j be the ball in the family $\{B(y_i, s_i)\}$.

(Stop until we can't find more) s.t: $A_j \cap (A_1 \cup \dots \cup A_{j-1}) = \emptyset$. with largest volumes

Then we get n -disjoint balls A_1, \dots, A_n .

$$B(x_1, r_1) \quad B(x_n, r_n)$$

For $x \in B(y_k, s_k)$, then we can find A_j s.t: $A_j \cap B(y_k, s_k) = \emptyset$.

Take the first $A_j = B(x_j, r_j)$. Then $r_j \geq s_k$. ($B(y_k, s_k) \cap (A_1 \cup \dots \cup A_{j-1}) = \emptyset$).

Then $B(y_k, s_k) \subset \underline{B(x_j, 3r_j)}$. metric property.

Since $K \subset \bigcup_K B(y_k, s_k) \subset \bigcup_j B(x_j, 3r_j)$

$$\Rightarrow m(K) \leq \sum_{j=1}^n m(B(x_j, 3r_j)) = (3^d) \sum_{j=1}^n m(B(x_j, r_j))$$

additional property
on Lebesgue measure

Rmk: The Vitali Lemma holds for Radon measure μ which satisfies "doubling property"

$$\mu(B(x, 2r)) \leq C \mu(B(x, r))$$

$\forall x, \forall r. \quad C \text{ is constant.}$

"Lebesgue differential" v.s. "Radon-Nikodym derivative"

$$= \lim_{r \rightarrow 0} \frac{\mu(B(x,r))}{m(B(x,r))} = f(x) \quad f = \frac{d\mu}{dm} \leftrightarrow d\mu = f dm. \quad (\mu \ll m)$$

& compact K. $\mu(K) < \infty$. $\mu(A) = \inf_{U \supset A} \mu(U)$

not required
in left side

Thm: Let μ be a locally finite, outer regular Borel measure on \mathbb{R}^d .

$$\text{Let } \mu = \mu_r + \mu_s. \quad \mu_r \ll m. \quad \mu_s \perp m \quad d\mu_r = f dm.$$

$$\text{Then for a.e. } x \in \mathbb{R}^d. \quad \lim_{r \rightarrow 0} \frac{\mu(B(x,r))}{m(B(x,r))} = \frac{d\mu_r}{dm}(x).$$

Proof: By Lebesgue differentiation theorem, for a.e. $x \in \mathbb{R}^d$

$$\lim_{r \rightarrow 0} \frac{\mu_r(B(x,r))}{m(B(x,r))} = \lim_{r \rightarrow 0} \frac{\int_{B(x,r)} f(t) dt}{m(B(x,r))} = f(x) = \frac{d\mu_r}{dm}(x).$$

It remains to prove

$$\boxed{\lim_{r \rightarrow 0} \frac{\mu_s(B(x,r))}{m(B(x,r))} = 0.}$$

Let $A \in \mathcal{B}$. s.t. $\mu_s(A) = 0$. $m(A^c) = 0$.

For $k \in \mathbb{N}$ let $E_k = \{x : \limsup_{r \rightarrow 0} \frac{\mu_s(B(x,r))}{m(B(x,r))} > \frac{1}{k}\} \cap A$

Need: $m(E_k) = 0$. $\forall k$.

By outer regularity. $\exists U_\varepsilon \supset A$ s.t. U_ε open.

$$\mu_r(U_\varepsilon) + \mu_s(U_\varepsilon) = \mu(U_\varepsilon) \leq \mu(A) + \varepsilon = \mu_r(A) + \varepsilon.$$

$$\Rightarrow \mu_s(U_\varepsilon) \leq \varepsilon.$$

For any $x \in E_k$. \exists open ball $B_x \subset U_\varepsilon$ s.t. $\mu_s(B_x) > \frac{1}{k} m(B_x)$.

Let $E_k \cap V_\varepsilon = \bigcup_{x \in E_k} B_x \subset U_\varepsilon$. Then $\forall c < m(V_\varepsilon)$

\exists finite balls B_{x_1}, \dots, B_{x_n} disjoint, s.t.

$$c < \sum_{j=1}^n m(B_{x_j}) \leq 3^d k \sum_{j=1}^n \mu_s(B_{x_j}) \leq 3^d k \mu_s(V_\varepsilon) \leq 3^d k \varepsilon.$$

$$\Rightarrow m(E_k) = 0$$

Both Lebesgue differential and the above hold if we replace

$\{B(x, r)\}$ by any family $\{E(x, r)\}$ that "shrink nicely" to each point. i.e. $x \in E(x, r)$. s.t.: ① $E(x, r) \subset B(x, r)$.
 ② $\exists \alpha \text{ indep of } x, r$
 s.t. $m(E(x, r)) > \alpha m(B(x, r))$,

Recall: $F \in C^1(R)$. F' is Riemann integral
 Lee 26 Differentiability of function $\int_a^b F'(x) dx = F(b) - F(a)$

1 The case of monotone function

Note: Consider $\mu_f = f dm$ $f \geq 0$ and f is integrable.

$$\text{Then } F(x) = \int_{(-\infty, x]} f dt = \mu_f((-\infty, x])$$

We have seen F' exists a.e. (Lebesgue differentiation)

In general, for any bounded measure μ .

We can define $F(x) = \mu((-\infty, x])$.

Example: $\mu = \chi_{[0,1]} dm + \delta_0 + \delta_1$

$$\sim F(x) = \begin{cases} 0 & x < 0 \\ 1 & 0 \leq x < 1 \\ 3 & x \geq 1 \end{cases} \quad F' = \chi_{[0,1]} \text{ a.e.}$$

Want: F . $\int_{(-\infty, x]} F'(t) dt = F(a) - F(b)$, let $F'(t) = f$ $F = \int_{(-\infty, x]} f(t) dt$

Note: These functions $F(x) = \mu((-\infty, x])$ are monotone increasing, right-continuous.

(Let $x_n \rightarrow x_0^+$ Then $(-\infty, x_n] \rightarrow (-\infty, x_0]$
 apply monotone convergence of measure)

Conversely, in PSet 2-2, for any increasing, right-continuous function $\tilde{F}: R \rightarrow R$, there exists a metric measure (\Rightarrow Borel measure) $\mu_{\tilde{F}}$ s.t. $\mu_{\tilde{F}}([a, b]) = \tilde{F}(b) - \tilde{F}(a)$
 (PSet-10-1, $\Rightarrow \mu_{\tilde{F}}$ is Radon \Rightarrow regular)

Then (Monotone differentiation theorem)

Any monotone function F on R is a.e. differentiable

Proof: WLOG we assume F is monotone increasing.

Let $G(x) = \bar{F}(x+)$. \Rightarrow right continuous.

(ii)

Observations: \bar{F} is discontinuous at most countable many points.

Reason: x_α discontinuous ($F(x_\alpha^-), F(x_\alpha^+)$) $\rightarrow \bigcup_\alpha (F(x_\alpha^-), F(x_\alpha^+))$
open-disjoint \rightarrow PSet₂₋₁₋₂. countable many
monotonically

$\Rightarrow G = F$ a.e.

$$\text{Consider } \lim_{h \rightarrow 0^+} \frac{G(x+h) - G(x)}{h} = \lim_{h \rightarrow 0^+} \frac{\mu_G((x, x+h])}{m((x, x+h])} = \frac{d(\mu_G)_r}{dm}(x) \text{ a.e. } x$$

$$\lim_{h \rightarrow 0^+} \frac{G(x) - G(x-h)}{h} = \lim_{h \rightarrow 0^+} \frac{\mu_G((x-h, x])}{m((x-h, x])} = \frac{d(\mu_G)_r}{dm}(x) \text{ a.e. } x$$

$\Rightarrow G'(x)$ exists for a.e. x .

Let $H = G - F$ We prove $H' = 0$ a.e.

Let $\{x_n\}$ be discontinuous points of F s.t. $H(x_n) \neq 0$.

and
at
the

By monotonicity, $H(x_n) > 0$. Moreover,

$$\sum_{x_n \in [-N, N]} H(x_n) \leq F(M) - F(-N)$$

$\Rightarrow \nu = \sum H(x_n) \delta_{x_n}$ is locally finite measure.
AND $\nu \perp m$.

By same argument $H'(x) = 0$ a.e. $\Rightarrow F' = G'$ a.e.

$$(F') = \lim_{h \rightarrow 0^+} \frac{F(x+h) - F(x)}{h} = \lim_{h \rightarrow 0^+} \frac{\mu_F((x, x+h])}{m((x, x+h])}$$

$$= \frac{d(\mu_F)_r}{dm}$$

Rough idea: Non-negative theory.

$$f \geq 0$$

μ measure

F monotone increasing

absolute integrable theory

$$f = f^+ - f^-$$

Signed measure $\mu = \mu_+ - \mu_-$

difference of two monotone increasing

which kind of F ?

Note: if F is a bounded increasing function. Then for any $x_1 < x_2 < \dots < x_n$, we have. $\sum_{i=1}^{n-1} |F(x_{i+1}) - F(x_i)| < F(+\infty) - F(+\infty)$

Def: The total variation of F is

$$\|F\|_{TV} = \sup \left\{ \sum_{i=1}^{n-1} |F(x_{i+1}) - F(x_i)| : \text{A n. } x_1 < x_2 < \dots < x_n \right\}.$$

We say F is a BV (bounded variation) function if $\|F\|_{TV} < +\infty$.

e.g. If F is monotone and bounded, then F is BV .
or the difference of two bounded and monotone functions

Rmk: If f is BV then F is bounded.

Thm // A function $F: R \rightarrow R$ is a BV function if and only if
it's the difference of two bounded monotone increasing functions
 $(\Rightarrow$ any BV function is a.e. differentiable)

Proof: Define "positive Variation function".

$$F^+(x) = \sup \left\{ \sum_{i=1}^{\infty} \max(F(x_{i+1}) - F(x_i), 0) : \text{A n. } x_1 < x_2 < \dots < x_n \leq x \right\}$$

Then $0 \leq F^+(x) \leq \|F\|_{TV}$, and $F^+(x) \nearrow$.

Let $F^-(x) = F^+(x) - F(x)$. Then F^- bounded and $F = F^+ - F^-$

Remains to prove: $F^-(x) \nearrow$.

Let $a < b$ Want: $F^-(a) \leq F^-(b)$; i.e.

$$F^+(b) \geq F^+(a) + F(b) - F(a). \quad (\#)$$

If $F(b) \leq F(a)$ $(\#)$ holds

If $F(b) > F(a)$ For any partition

$$\begin{cases} x_1 < x_2 < \dots < x_n \leq a \\ x_1 < x_2 < \dots < x_h < x_{h+1} = a < x_{h+2} = b \end{cases}$$

By def of F^+ , $F^+(b) - F^+(a) \geq F(b) - F(a)$.

It locally BV f. f is a.e. differentiable and $f' \in L^1_{loc}(k)$
WANT $\int_a^b F'(x) dx = F(b) - F(a) \quad (\#)$

Note: Even for monotone function $(\#)$ may fail.
Note Generally: $F(x) = F(a) + \int_a^x F'(t) dt$.

In Lec 11, Lebesgue integral is absolutely continuous

$$f \in L^1 \Rightarrow \forall \varepsilon > 0 \exists \delta > 0 \text{ s.t. } \forall A \text{ with } m(A) < \delta.$$

$$\text{we have } \int_A |f| dx < \varepsilon.$$

$$\text{Take } A = \bigcup_{i=1}^n (a_i, b_i) \text{ (disjoint)} \quad \Rightarrow \quad \left| \int_A f \right| \leq \sum_{i=1}^n \left| \int_{(a_i, b_i)} f(t) dt \right| = \sum_{i=1}^n (F(b_i) - F(a_i)) \\ \leq \sum_{i=1}^n |f(b_i) - f(a_i)| < \varepsilon.$$

Def: // A function f is absolutely continuous (AC). $\forall \varepsilon \exists \delta$.
 If for any disjoint (a_i, b_i) $1 \leq i \leq n$ with $\sum_{i=1}^n (b_i - a_i) < \delta$.
 we have $\sum_{i=1}^n |f(b_i) - f(a_i)| < \varepsilon$.

Example: $f \in L^1 \Rightarrow F(x) = \int_{(-\infty, x]} f(t) dt$ is AC.

Prop 1 // If F is AC on $[a, b]$. Then F is BV.

Prop 2 // If F is AC on $[a, b]$ and $F'(x) = 0$ a.e. then $F = \text{constant}$
 Only need to prove $F(a) = F(b)$.

Proof: $F'(x) = 0 \Rightarrow \exists (a_i, b_i) \ni x$. s.t. $|F(b_i) - F(a_i)| \leq \varepsilon (b_i - a_i)$
 can take (a_i, b_i) arbitrarily small

$$\text{Let } A = \{x | F'(x) = 0\} \Rightarrow m(A) = b - a.$$

Vitali covering of $A \Rightarrow \exists (a_i, b_i)$ s.t. $\sum_{i=1}^M |b_i - a_i| > (b - a) - \delta$
 $[a, b] \setminus \bigcup_{i=1}^M (a_i, b_i) = \bigcup_{k=1}^N (\alpha_k, \beta_k) \rightsquigarrow \sum_{k=1}^N |\beta_k - \alpha_k| < \delta$.

$$\text{Finally, } |F(b) - F(a)| \leq \varepsilon. \text{ Const.}$$

Prop 3 // Let F be a monotone increasing function
 Then $\int_{[a,b]} F' x dx \leq F(b) - F(a)$.

In particular $F' \in L^1([a, b])$.

$$\text{LHS} = \int_{[a,b]} \lim_{h \rightarrow 0} \frac{F(x + \frac{1}{h}) - F(x)}{\frac{1}{h}} dx.$$

$$\leq \liminf_{h \rightarrow 0} \int_{[a,b]} \frac{F(x + \frac{1}{h}) - F(x)}{\frac{1}{h}} dx.$$

$$\text{Factor} = \liminf_{h \rightarrow 0} \frac{1}{h} \left[\int_{[b, b + \frac{1}{h}]} F(x) dx - \int_{[a, a + \frac{1}{h}]} F(x) dx \right] \rightarrow F(b) - F(a)$$

The fundamental theorem of calculus (Version 2)

Let $F: [a, b] \rightarrow \mathbb{R}$ be AC. Then $\int_{[a, b]} F'(x) dt = F(b) - F(a)$

Proof: $AC \Rightarrow BV \Rightarrow F' \text{ exists a.e. } F(x) = F(a) + \int_{[a, x]} F'(t) dt$

Moreover, $F: sBV \Rightarrow F = F_1 - F_2$, F_i bounded monotone
 $\Rightarrow F' \in L^1([a, b])$.

Let $G(x) = \int_{[a, x]} F'(t) dt + F(a)$.

Then by a.c. of Lebesgue integral. we know G is AC.
 $\Rightarrow F - G$ is AC.

By Fundamental Theorem of Calculus (Version 1).

we know $G'(x) = F'(x)$ a.e.

$$\Rightarrow (F - G)' = 0 \text{ a.e.}$$

By Prop 2 $F - G = \text{const}$, $F(a) = G(a) \Rightarrow F = G$.

Rmk: $AC \Rightarrow F(x) = F(a) + \int_a^x F'(t) dt$
 \Leftarrow If and $F' \in L^1$

Lec. 27.

1 Some applications

Theorem (Integration by parts) Let f, g be AC functions on $[a, b]$. Then $\int_{[a,b]} f(t)g(t)dt = \left[f(t)g(t) \right]_a^b - \int_{[a,b]} f'(t)g'(t)dt$

Proof. By Pset 14-1-3 f, g is AC. Since

$$(fg)' = f'g + fg' \text{ a.e.}$$

$$\Rightarrow f(b)g(b) - f(a)g(a) = \int_{[a,b]} (fg)' dt = \int_{[a,b]} f'g dt = \int_{[a,b]} fg' dt.$$

Theorem Let f be AC function on $[a, b]$. Then $\|f\|_{TV([a,b])} = \int_{[a,b]} |f'(t)|dt = \|f'\|_{L^1([a,b])}$.

Proof: For any partition $a = x_1 < x_2 < \dots < x_n = b$, we have

$$\sum_{i=1}^{n-1} |f(x_{i+1}) - f(x_i)| = \sum_{i=1}^{n-1} \left| \int_{[x_i, x_{i+1}]} f'(t)dt \right| \leq \sum_{i=1}^{n-1} \int_{[x_i, x_{i+1}]} |f'(t)| dt = \int_{[a,b]} |f'(t)| dt = \|f'\|_{L^1([a,b])}$$

Want: $\|f'\|_{L^1} \leq \|f\|_{TV} + \varepsilon$

By Pset 6-1-2. $\because f' \in L^1([a,b])$. \exists step function g on $[a, b]$ s.t. $\|h\|_{L^1} < \varepsilon$, where $h = f' - g$.

$$\text{Let } G(x) = \int_{[a,x]} g(t)dt \quad H(x) = \int_{[a,x]} h(t)dt.$$

$$\Rightarrow f(x) = f(a) + \int_{[a,x]} f'(t)dt = f(a) + G(x) + H(x)$$

Since $H' = h \in L^1$. H is AC. $\Rightarrow \|H\|_{TV} \leq \|h\|_{L^1} < \varepsilon$.

By triangle $\|f\|_{TV} \geq \|G\|_{TV} - \|H\|_{TV} \geq \|G\|_{TV} - \varepsilon \geq \|f'\|_{L^1} - 2\varepsilon$

Write $g = \sum_{i=1}^n c_i \chi_{[x_i, x_{i+1}]}$, where $a = x_1 < x_2 < \dots < x_n = b$.

$$\begin{aligned} \Rightarrow \|G\|_{TV} &\geq \sum_{i=1}^n |G(x_{i+1}) - G(x_i)| = \sum_{i=1}^n \left| \int_{[x_i, x_{i+1}]} c_i dt \right| \\ &= \int_{[a,b]} \sum_{i=1}^n |c_i| \chi_{[x_i, x_{i+1}]} dt = \|g\|_{L^1}. \end{aligned}$$

Rank: These theorem holds for c -valued functions.

A geometric application: rectifiable curves.

2. Lipschitz Continuity

Def: A function $f: \mathbb{R} \rightarrow \mathbb{R}$ is Lipschitz continuous if $\exists M > 0$. s.t. $|f(x) - f(y)| \leq M|x - y| \quad \forall x, y$.

→ Lipschitz continuous function between metric space
 $f: \mathbb{R}^d \rightarrow \mathbb{R}^n \quad |f(x_1) - f(x_2)| \leq M|x_1 - x_2|$

Def: $\text{Lip}(f) = \sup_{x \neq y} \frac{|f(x) - f(y)|}{|x - y|}$

Prop: If f is AC. and $|f'| \leq M$ a.e. then f is Lipschitz continuous, and $\text{Lip}(f) \leq M$.

$$\begin{aligned} |f(x) - f(y)| &= \left| \int_{[x,y]} f'(t) dt \right| \\ &\leq \int_{[x,y]} |f'(t)| dt \leq M|x - y|. \end{aligned}$$

Thm: Let $F: [a,b] \rightarrow \mathbb{R}$ be Lipschitz continuous with $\text{Lip}(F) = M$. Then $\exists f \in L^\infty([a,b]) \subset C'([a,b])$ with $\|f\|_{C^\infty} \leq M$ s.t. $F(x) - F(a) = \int_{[ax]} f(t) dt$

Cor: If $F: \mathbb{R} \rightarrow \mathbb{R}$ Lipschitz continuous, $\text{Lip}(F) = M$. then F is AC, and $|F'| \leq M$ a.e.

Cor: Any Lipschitz continuous function $F: \mathbb{R} \rightarrow \mathbb{R}$ is a.e. differentiable

Proof: We define a linear functional on the space of step functions by $\langle \left(\sum_{i=1}^n a_i \chi_{[x_i, x_{i+1}]} \right) \rangle = \sum_{i=1}^n a_i (F(x_{i+1}) - F(x_i))$

For any $h = \sum_{i=1}^n a_i \chi_{[x_i, x_{i+1}]}$ (step function) we have

$$|L(h)| = \left| \sum_{i=1}^n a_i (F(x_{i+1}) - F(x_i)) \right| \leq \sum_{i=1}^n |a_i| \cdot M|x_{i+1} - x_i|$$

$$= M \|h\|_{L^1([a,b])}.$$

$\Rightarrow L$ is bounded on the space of step functions which is dense in $L^1([a,b])$

$\rightarrow L$ extends a continuous linear functional on $L^1([a,b])$.

By Riesz representation $\exists f \in L^\infty([a,b])$, s.t. $L(g) = \int_{[a,b]} f(t) g(t) dt$.

Take $g = \chi_{[a,x]}$ $x \in [a,b]$.

$$\text{we get } F(x) - F(a) = L(g) = \int_{[a,b]} f(t) \chi_{[a,x]}(t) dt$$

$$= \int_{[a,x]} f(t) dt.$$

Now consider functions $f: \mathbb{R}^d \rightarrow \mathbb{R}$

Recall: Directional derivative of f in the direction

$$v \in \mathbb{R}^d \text{ at } x_0, D_v f(x_0) = \lim_{h \rightarrow 0} \frac{f(x_0 + hv) - f(x_0)}{h}$$

f is differentiable at x_0 if \exists a linear map $L: \mathbb{R}^d \rightarrow \mathbb{R}$
 s.t. $\lim_{h \rightarrow 0} \frac{|f(x_0 + h) - f(x_0) - L(h)|}{\|h\|} = 0$. \uparrow depends on x_0 .

Rmk: If f is differentiable, then $D_v f(x_0) = L(v)$.

In this case we can denote $L(e_i) = D_{e_i} f(x_0) = \frac{\partial f}{\partial x_i}(x_0) = \partial_i f$.

Then we write $\nabla f = (\partial_1 f, \dots, \partial_d f)$
 $\Rightarrow L(v) = \nabla f \cdot v$

It's possible that all $D_v f(x_0)$ exists, but f is not differentiable at x_0 . [One needs: continuity of partial derivatives].

Theorem (Rademacher differentiability theorem) Let $F: \mathbb{R}^d \rightarrow \mathbb{R}^n$ be a Lipschitz continuous function, then F is a.e. differentiable.

Proof: Since F is continuous, $D_v F(x_0)$ exists if and only if

$$\limsup_{\substack{\text{Q} \\ h \rightarrow 0}} \frac{F(x_0 + hv) - F(x_0)}{h} = \liminf_{\substack{\text{Q} \\ h \rightarrow 0}} \frac{F(x_0 + hv) - F(x_0)}{h}.$$

$\Rightarrow \forall v \in \mathbb{R}^d$ the set

$E_v := \{x \in \mathbb{R}^d : D_v F(x_0) \text{ does not exist}\}$ is a Borel set

Moreover, $D_v F$ is a measurable function on E_v^c and $|D_v F| \leq c$.

Fact: $m(E_v) = 0$

Proof: Observe for $v=0$.

For $v \neq 0$ apply a linear transformation $v \mapsto e_1$.

So wlog, we need to show $D_1 F$ exists a.e.

We split $\mathbb{R}^d = \mathbb{R} \times \mathbb{R}^{d-1}$

Fix any (x^0, y^0)

$x^0 \quad y^0$

$D_1 F(x^0, y^0)$ exists $\Leftrightarrow \underline{x \mapsto F(x, y^0)}$ is differentiable at x^0 .

$\Rightarrow E_{y^0} = \{x^0 \in \mathbb{R} : (x^0, y^0) \in E_y\}$ is a null set.

By Fabini (Tonelli).

$$m(E_v) = \iint \chi_{E_v} dx dy = \underbrace{\int_{\mathbb{R}^{d-1}} \int_R \chi_{E^{y^0}(x)} dx dy}_{y^0} = 0.$$

$\Rightarrow m(\bigvee_{v \in \mathbb{Q}^d} E_v) = 0$.

i.e. For a.e. $x \in \mathbb{R}^d$, $D_v F(x)$ exists for all $v \in \mathbb{Q}^d$.

Let $g \in C_c^\infty(\mathbb{R}^d)$ Then

$$\begin{aligned} \int_{\mathbb{R}^d} D_v F(x) g(x) dx &= \int_{\mathbb{R}^d} \lim_{\substack{\text{Q} \\ h \rightarrow 0}} \frac{F(x+hv) - F(x)}{h} g(x) dx \\ &\stackrel{(LCT)}{=} \lim_{\substack{\text{Q} \\ h \rightarrow 0}} \int_{\mathbb{R}^d} \frac{F(x+hv) - F(x)}{h} g(x) dx \end{aligned}$$

$$= \lim_{\delta \rightarrow 0} \int_{R^d} f(x) \underbrace{\frac{g(x+hv) - g(x)}{h}}_{\text{compact supp.}} dx \xrightarrow{\text{DCT}} \int_{R^d} f(x) D_v g(x) dx.$$

\Rightarrow LHS is linear in v

So if we let $v = \sum_{i=1}^d v_i e_i$ then

$$\int (D_v f(x)) g(x) dx = \sum v_i \int (e_i f)(x) g(x) dx.$$

$$\Rightarrow \int_{R^d} (D_v F - v \cdot \nabla F) g(x) dx = 0. \quad \forall g \in C_0^\infty(R^d)$$

$$\Rightarrow D_v F = v \cdot \nabla F \text{ a.e. } x. \quad \forall v \in \mathbb{Q}^d.$$

Let $A = \{x \in R^d : D_v F(x) = v \cdot \nabla F(x), \forall v \in \mathbb{Q}^d\}$.

Claim: F is differentiable on A .

Define $F(h) = F(x_0 + h) - F(x_0) - h \cdot \nabla F(x_0)$ $\forall h \in A$.

Want $\lim_{h \rightarrow 0} \frac{|\tilde{F}(h)|}{|h|} = 0$. $\uparrow \tilde{F}(0) = 0$.
 \tilde{F} is Lipschitz

Let $\varepsilon > 0$, $h = rv$, $r = |h|$, $v \in S^{d-1}$

Take a finite set $V_\varepsilon \subset \mathbb{Q}^d \cap S^{d-1}$, s.t. $\exists v \in V_\varepsilon$.

with $|u-v| < \varepsilon$.

$$\Rightarrow \frac{|\tilde{F}(rv) - \tilde{F}(r0)|}{r} < \varepsilon \quad \forall v \in V_\varepsilon \quad r \text{ small.}$$

$$\Rightarrow |\tilde{F}(rv)| \leq r\varepsilon.$$

Since \tilde{F} is Lipschitz

$$\Rightarrow |\tilde{F}(h) - \tilde{F}(rv)| = |\tilde{F}(rv) - \tilde{F}(rv)| \leq \text{Lip}(\tilde{F}) \cdot |h| \cdot |u-v| \leq \text{Lip}(\tilde{F}) \cdot \varepsilon r$$

$$\Rightarrow \lim_{h \rightarrow 0} \frac{|\tilde{F}(h)|}{|h|} = 0.$$

Pset 14-1.

$$1. \text{ ④ } \forall a = y_0 < \dots < x_{n+1} = b = y_n < \dots < y_{m+1} = c.$$

$$\|F\|_{TV([a,c])} \geq \sum |F(x_{k+1}) - F(x_k)| + \sum |F(y_{k+1}) - F(y_k)|$$

$$\text{左边取 sup. } \Rightarrow \|F\|_{TV([a,c])} \geq \|F\|_{TV([a,b])} + \|F\|_{TV([b,c])}.$$

$$\text{另一方面 } \forall \varepsilon > 0 \exists a \leq x_0 < \dots < x_{n+1} \leq b < y_0 < \dots < y_{m+1} \leq c.$$

$$\begin{aligned} \|F\|_{TV([a,c])} - \varepsilon &\leq \sum |F(x_{k+1}) - F(x_k)| + \sum |F(y_{k+1}) - F(y_k)| + |y_0 - x_{n+1}| \\ &\leq \sum |F(x_{k+1}) - F(x_k)| + |b - x_{n+1}| \\ &\quad + \sum |F(y_{k+1}) - F(y_k)| + |y_0 - b| \\ &\leq \|F\|_{TV([a,b])} + \|F\|_{TV([b,c])} \end{aligned}$$

$$2. \text{ ① } D^+F = \limsup_{h \rightarrow 0} \frac{F(x+h) - F(x)}{h}$$

$$G = \limsup_{\substack{h \rightarrow 0 \\ h \neq 0}} \frac{F(x+h) - F(x)}{h}$$

$$\forall \varepsilon > 0 \exists D^+F = G. \quad \forall h > 0 \quad \forall h_1 < h < h_2, h_1, h_2 \in \mathbb{Q}.$$

$$\frac{h_1}{h} > 1-\varepsilon, \quad \frac{h_2}{h} < 1+\varepsilon.$$

$$\text{由 } \left(\frac{h_1}{h} \right) \frac{F(x+h) - F(x)}{h_1} \leq \frac{F(x+h_1) - F(x)}{h} \leq \frac{F(x+h) - F(x)}{h} \leq \frac{F(x+h_2) - F(x)}{h_2} \left(1+\varepsilon \right)$$

$$\text{取 } \limsup_{h \rightarrow 0} \Rightarrow (1-\varepsilon)G(x) \leq D^+F(x) \leq (1+\varepsilon)G(x).$$

$$② F \nearrow \Rightarrow D^+F = F' \text{ a.e.}$$

$$m(\{D^+F > \alpha\}) = m(\{F' > \alpha\}) \leq \frac{1}{\alpha} \int_a^b F' dx \leq \frac{1}{\alpha} (F(b) - F(a))$$

$$3. \text{ ③ } \forall \varepsilon > 0 \exists \delta > 0 \exists [a_i, b_i] A \subset \bigcup [a_i, b_i] \sum (b_i - a_i) < \delta. \quad \boxed{\|F(b_i) - F(a_i)\| < \varepsilon}$$

$$\text{由 } a_i' = \arg \min_{x \in [a_i, b_i]} F(x) \quad b_i' = \arg \max_{x \in [a_i, b_i]} F(x) \quad \boxed{\|b_i' - a_i'\| < \varepsilon}$$

$$\text{由 } \boxed{[a_i', b_i'] \subset [a_i, b_i]}, \quad F([a_i, b_i]) = [F(a_i'), F(b_i')]$$

$$m(F(A)) \leq \sum m(F[a_i, b_i]) \leq \sum |F(b_i') - F(a_i')| < \varepsilon.$$

$$4. F_j \geq 0, F_j \nearrow, F = \sum F_j < \infty. \text{ pf. } F' = \sum_{j=0}^{\infty} F'_j.$$

$$\text{pf. 不妨设 } F_j \text{ 在 } x \text{ 连续. } F_j(x) = F_j(x+) \quad F(x) = F(x+)$$

$$\mu_F = \sum \mu_{F_j} \quad \text{设 } \mu \text{ ac. } (\mu)_s \quad \begin{matrix} \downarrow \\ \ll_m \end{matrix} \quad \begin{matrix} \downarrow \\ \perp_m \end{matrix}$$

$$\mu_F = \sum \mu_{F_j} = \sum_{i < m} (\mu_{F_i})_{ac} + (\mu_{F_i})_s \quad \text{由 分解定理} - \quad (\mu_F)_{ac} = \sum (\mu_{F_j})_{ac} \\ \forall A \in \mathcal{B}(R).$$

$$(\mu_F)_{ac}(A) = \sum (\mu_{F_j})_{ac}(A) = \sum \int_A F'_j dm \xrightarrow{\text{MCIT}} \int_A \sum F'_j dm. \\ = \int_A F' dm \quad \Rightarrow \quad F = \sum F'_j$$

PSet 14-2.

i) $\{u_\alpha\}_{\alpha \in \mathbb{Z}} \text{ Lip. } \text{Lip}(u_\alpha) \leq L.$

ii) $U = \sup u_\alpha. \quad \forall x_0 \in X. \quad \text{pf: } \text{Lip}(U) \leq L.$

iii) $U = \inf u_\alpha. \quad \text{pf: } \text{Lip}(U) \leq L.$

Pf. ii) $\forall x \in X. \quad \forall \alpha$

$$|u_\alpha(x) - u_\alpha(x_0)| \leq L \cdot d(x, x_0)$$

$$U(x) = \sup_\alpha u_\alpha(x) \leq \sup_\alpha u_\alpha(x_0) + L \cdot d(x, x_0) < \infty.$$

$\Rightarrow U$ well-defined

$$U(x) = \sup_\alpha u_\alpha(x) \leq \sup_\alpha (u_\alpha(x) - u_\alpha(y)) + \sup_\alpha u_\alpha(y). \\ \leq L \cdot d(x, y) + U(y).$$

2. $A \subset X. \quad \forall \text{Lip } f: A \rightarrow \mathbb{R}, \quad \text{Lip}(f) = L \quad \text{令 } \tilde{f}(x) = \sup_{y \in A} (f(y) + L \cdot d(x, y))$

Proof: $\text{Lip}(\tilde{f}) \leq L.$

Pf. $\forall y \in A. \quad \tilde{f}(y) = f(y) + L \cdot d(x, y).$

$$|\tilde{f}(y) - \tilde{f}(x)| = |f(y) + L \cdot d(x, y) - f(x) - L \cdot d(x, y)| \leq L \cdot d(x, y) \\ \Rightarrow \text{Lip}(\tilde{f}) \leq L.$$

$\forall x_0. \quad |\tilde{f}(y) - \tilde{f}(x_0)| \leq L \cdot d(x_0, y).$

$$\Rightarrow \tilde{f}(x_0) \geq f(y) + L \cdot d(x_0, y) \geq f(x_0) > \infty. \quad \text{由 } \bar{y} \notin A.$$

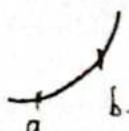
③ $f \in \mathcal{C}^1$ Prove.

① f 可微 a.e.

② f' 1.

③ $\forall a < x < b. \quad x = (1-t)a + tb$

$$\Rightarrow \frac{f(x) - f(a)}{x - a} \leq \frac{f(b) - f(a)}{b - a} \leq \frac{f(b) - f(x)}{b - x}$$



④ 用 LDT 记 $\int_{R^d} h \varphi = 0$ for $\forall \varphi \in C_0^\infty$. 且 $h = 0$ a.e.

$$h \in L^1_{loc}(R^d)$$

且 ν signed, $\mu \perp \nu \Rightarrow |\mu| \perp \nu$.

(反证) 若 $\mu \mid_{X_2} = 0$ 但 $|\mu| \mid_{X_2} \neq 0$. $\exists A \subset X_2$ s.t:

~~且~~ $\mu(A) > 0$. $X = X_+ \cup X_-$. $\mu_+|_{X_-} = 0$, $\mu|_{X_+} = 0$.

$$\underline{\mu_+(A \cap X_+) + \mu(A \cap X_-) > 0}$$

$$A \subset X_2. A \cap X_+ \subset X_2. \mu(A \cap X_+) = \mu_+(A \cap X_+) > 0.$$