

Lecture 1 Motivation for measure theory

1.1

1.i) Syllabus + Introduction

1.ii) Motivation for Real Analysis

Real analysis is the rigorous foundation of calculus, convergence, and functions.

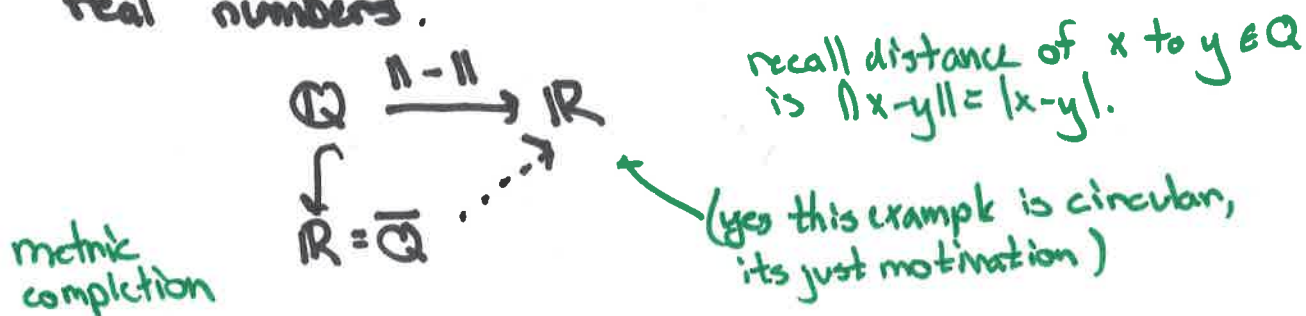
Applications 1.1 : Applications of real analysis include

- 1) Description of physical systems by ODEs and dynamical systems
- 2) Study of partial differential equations (PDEs) describing Electromagnetism, heat, waves, quantum particles, quantum fields, fluids, plasmas, ..
- 3) Analytic Number theory + Automorphic forms
- 4) Representation theory and harmonic analysis
- 5) Probability theory
- 6) Optimization problems in physics, math, economics, etc.
- 7) Study of fractals and chaotic phenomena
- 8) Dynamics and ergodic theory

(Informal) Def 1.2: A space X is said to be complete if limits of $x_i \in X$ converge to points of X .

Ex 1.3: The rational numbers $\mathbb{Q}$ are an incomplete (metric) space, i.e. not all Cauchy sequences converge.

To fix this is the main motivation for introducing real numbers.



→ in $\mathbb{R}$ every Cauchy sequence converges.

Motivation 1.9 (Fourier Series)

(1.3)

Fourier Series gives a way of expressing a ^(periodic) function as a sum

$$f(x) = \sum_{n=0}^{\infty} a_n \cos(nx) + b_n \sin(nx).$$

What does it mean for this to converge? $f_N = \sum_{n=0}^N$ then

$f_N(x) \rightarrow f(x)$ for every x (pointwise convergence)

$\sup |f_N(x) - f(x)| \rightarrow 0$ (uniform convergence)

$\int |f_N(x) - f(x)| \rightarrow 0$ (Integral convergence).

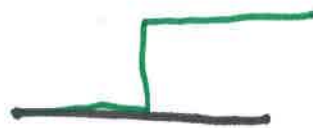
When are each true? For what sequences $\{a_n, b_n\}$ is $f(x)$ continuous? Differentiable.

Studying these questions in the context of PDEs was historically a main motivation for measure theory.

Motivation 1.4: We can define a metric on the space of continuous functions $C^0[a,b]$ on an interval $[a,b] \subseteq \mathbb{R}$ by $\|f-g\| = d(f,g) = \int_a^b |f-g| dx$ (Riemann integral)

But this space is certainly not complete!

Ex 1.5:



(not continuous, but is integrable)

Difference of integrals $\leq \frac{1}{2}bh \rightarrow 0$.

Ex 1.6: Consider $F(x) = \begin{cases} 1 & x \in \mathbb{Q} \\ 0 & x \in \mathbb{R} \setminus \mathbb{Q} \end{cases}$.

Then $F(x)$ is not Riemann integrable! But it is a limit of these (more next time).

Question 1.7: Can we find a space $\bar{X}$ such that

$$C^0[a,b] \subseteq RI[a,b] \xrightarrow{\int_a^b f dx} \mathbb{R}$$



is the metric completion of Riemann integrable functions?

A: Yes! It is $L^1[a,b]$ the Lebesgue integrable functions.

Motivation 1.8: If f_n is a sequence of functions, when

$$\text{does } \lim_{n \rightarrow \infty} \int_a^b f_n(x) dx = \int_a^b \lim_{n \rightarrow \infty} f_n(x) dx ?$$

This is very annoying w/ Riemann (and RIG may not exist)

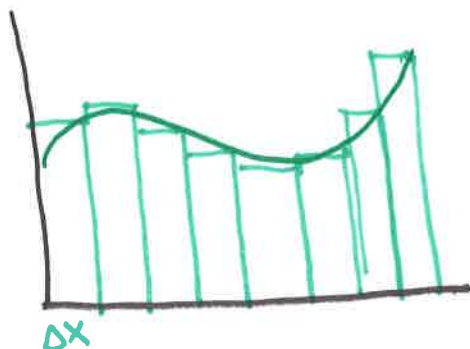
It becomes easy with Lebesgue Dominated Convergence Theorem (Lecture 9).

Lecture 2 σ -algebras and measures

2.1

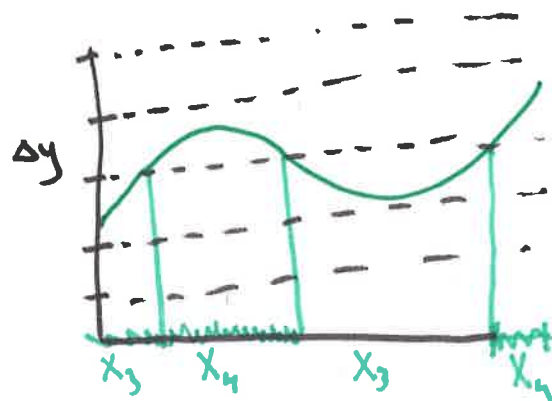
Ex 2.1 : Coin counting example.

2.1) Intuition for the Lebesgue Integral



Divide up x-axis, then

$$\int_0^1 f dx = \lim_{\Delta x \rightarrow 0} \sum_{i=1}^N \Delta x_i \cdot f(x_i)$$



$$\int_0^1 f dx = \lim_{\Delta y \rightarrow 0} \sum_{j=1}^N j \Delta y \cdot m(X_j)$$

"size or measure of X_j "

Goal 2.2 : Define a measure

$$m : \{A \mid A \subseteq \mathbb{R}^n \text{ is a subset}\} \rightarrow \mathbb{R} \cup \{\infty\}$$

that satisfies "reasonable" properties.

- 1) $m([a, b]) = b - a$ (normalization)
- 2) $m(A \cup B) = m(A) + m(B) - m(A \cap B)$ (well-behaved under $\cup, \cap$)
- 3) $m(A + x) = m\{a + x \mid a \in A\} = m(A)$ (translation invariance)

Rem 2.3 : $\nexists m$ satisfying the above for all subsets

$$\left\{ \begin{array}{l} \text{finite unions} \\ \text{of intervals} \end{array} \right\} \subsetneq \left\{ \begin{array}{l} \text{Borel} \\ \text{sets} \end{array} \right\} \subsetneq \left\{ \begin{array}{l} \text{measurable} \\ \text{subsets} \end{array} \right\} \subsetneq \left\{ \begin{array}{l} \text{All subsets} \end{array} \right\}$$

$\downarrow m$
 $\mathbb{R} \cup \{\infty\}$

Rem 2.12 : Every $x \in \mathbb{R}$ has a continued fraction expression

(2.4)

$$x = a_0 + \frac{1}{a_1 + \frac{1}{a_2 + \frac{1}{\dots}}}$$

wl $a_0 \in \mathbb{Z}$ and $a_i \in \mathbb{N}, \mathbb{Z}^{\geq 0}$

$K = \left\{ x \in \mathbb{R} \mid x \notin \mathbb{Q} \text{ and } (a_0, a_1, \dots) \text{ has a subsequence } a_{i_k} \right. \\ \left. \text{st } \{a_{i_k} \mid a_{i_{k+1}}\} \right\}$

is not a Borel set.

2.iii) Measures :

Def 2.13 : Let (X, Σ) be a metric space with σ -algebra.

A measure

$$\mu : \Sigma \longrightarrow \mathbb{R} \cup \{\infty\}$$

is a function such that

i) $\mu(E) \geq 0 \quad \forall E \in \Sigma$

ii) $\mu(\emptyset) = 0$

iii) If $\{E_i\}_{i=1}^{\infty}$ are a countable disjoint collection

$$\mu\left(\bigcup_{i=1}^{\infty} E_i\right) = \sum_{i=1}^{\infty} \mu(E_i).$$

Def 2.14 : A measure space is a triple (X, Σ, μ)

of a metric space and σ -algebra with a measure.

It is complete if $E \subseteq A \in \Sigma$ and then $\mu(A) = 0 \Rightarrow \mu(E) = 0$.

Theorem 2.15 : There exists a measure m and σ -algebra $\mathcal{M}(\mathbb{R}^n)$

making $(\mathbb{R}^n, \mathcal{M}(\mathbb{R}^n), m)$ into a measure space with the addition properties

i) $m([a, b]) = b - a$

ii) m is complete

iii) If $A \in \mathcal{M}(\mathbb{R}^n)$, $A+x \in \mathcal{M}(\mathbb{R}^n)$ and $m(A) = m(A+x)$.

iv) $\mathcal{B}(\mathbb{R}^n) \subseteq \mathcal{M}(\mathbb{R}^n)$

m is called the Lebesgue measure, and $\mathcal{M}(\mathbb{R}^n)$ the measurable sets.

Note I_x, I_y distinct or equal. If $x \neq y$, then $I_x \cup I_y$ is an interval contradicting maximality unless $I_x = I_y$. And each I_x is open so contains some $q_i \in \mathbb{Q}$. Since the collection of distinct intervals contains distinct rationals, it is countable. $\square$

Lemma 2.8: Every open $U \subseteq \mathbb{R}^d$ is a union

$$U = \bigcup_{n=1}^{\infty} [a_n, b_n] \times \dots \times [a_n, b_n]$$

that have disjoint interiors.

Proof: Exercise (see pages 7-8).

Let $P(X)$ denote the power set $P(X) = \{A \mid A \subseteq X\}$.

Definition 2.9: A σ -algebra on X is a subset $\Sigma \subseteq P(X)$

such that

i) $X \in \Sigma$

ii) $A \in \Sigma \Rightarrow A^c \in \Sigma$ (complements)

iii) $A_i \in \Sigma \quad i=1, \dots \Rightarrow \bigcup_{i=1}^{\infty} A_i \in \Sigma$ (countable unions).

(same for intersections by ii) $\cap A_i = (\cup A_i^c)^c$.

Def 2.10: If X is a metric space, the σ -algebra

$$\mathcal{B}(X) = \{ \text{smallest } \sigma\text{-algebra containing open sets} \}$$

$$= \bigcap_{\Sigma \text{ open } \subseteq \Sigma} \Sigma \quad \leftarrow P(X) \text{ is a } \sigma\text{-algebra so non-empty.}$$

is called the Borel sets of X .

Rem 2.11:

$\left\{ \begin{array}{l} \text{open/closed} \\ \text{sets} \end{array} \right\} \subsetneq \left\{ \begin{array}{l} \text{Borel} \\ \text{subsets} \end{array} \right\}$

$\uparrow$ $[0, 1/2) = \bigcup \text{closed}$

$\nwarrow$ has cardinality of $\mathbb{R}$

$\nearrow$ bigger cardinality "most sets are not Borel".

$\neq P(\mathbb{R})$

Ex 2.4 : Consider $f: [0,1] \rightarrow \mathbb{R}$

[2.2]

$$f(x) = \begin{cases} 1 & x \in \mathbb{Q} \cap [0,1] \\ 0 & \text{else} \end{cases}$$



Claim : $f(x)$ is not Riemann integrable

$$\left| \sum_{i=1}^N \inf_{x \in [x_{i-1}, x_i]} f(x) \cdot \Delta x - \sum_{i=1}^N \sup_{x \in [x_{i-1}, x_i]} f(x) \cdot \Delta x \right| = 1$$

For all $\Delta x > 0$.

But in a Lebesgue sense

$$\int_0^1 f(x) dx = 1 \cdot \underbrace{m(\mathbb{Q} \cap [0,1])}_{=0}$$

$\forall \varepsilon > 0$, $m(\mathbb{Q} \cap [0,1]) \leq \varepsilon$. Let $\{q_n\}_{n \in \mathbb{N}}$ be an enumeration of $\mathbb{Q}$.

Let $I_n = [q_n - \frac{\varepsilon}{2^{n+1}}, q_n + \frac{\varepsilon}{2^{n+1}}]$. Then $\mathbb{Q} \cap [0,1] \subseteq \bigcup I_n$

but $m(\bigcup I_n) \leq \sum_{n=1}^{\infty} m(I_n) = m(\cap \dots) \leq \sum_{n=1}^{\infty} m(I_n) = \varepsilon \sum_{n=1}^{\infty} \frac{1}{2^n} = \varepsilon \cdot 1 = \varepsilon$.

2.ii) Open and Borel sets, σ -algebras.

Def 2.5 : An open subset $U \subseteq \mathbb{R}$ is one such that $\forall x \in U, \exists r_x$ st $B_{r_x}(x) \subseteq U$.

A closed subset C is one so that $C^c = \{x \in \mathbb{R} \mid x \notin C\}$ is open.

Facts 2.6 : 1) Arbitrary unions of open sets are open (inf of closed are closed)
2) finite intersection of open sets is open ($\bigcup$ of closed sets is closed).

Lemma 2.7 : Every open $U \subseteq \mathbb{R}$ may be written

$$U = \bigcup_{n=1}^{\infty} (a_n, b_n)$$

where $a_n, b_n = \pm\infty$ is allowed.

Proof : $\forall x \in U$, let $I_x = \left(\inf_{a \in U} a, \sup_{b \in U} b \right)$
st $(a, x) \subseteq U$ st $(x, b) \subseteq U$

$I_x \subseteq U$
 I_x is open, and $x \in I_x$, so $U = \bigcup I_x$.

Lecture 3 Outer Measure + Lebesgue Measure

3.1

Recall: We wish to construct a measure ("Lebesgue measure") and a σ -algebra $\mathcal{M}(\mathbb{R}^n)$ such that

- i) $m(Q) = |Q|$ if $Q = [a_1, b_1] \times \dots \times [a_n, b_n]$ is a cube
- ii) m is complete and translation invariant
- iii) $\mathcal{B}(\mathbb{R}^n) \subseteq \mathcal{M}(\mathbb{R}^n)$

Ex 3.1: It is easy to construct an arbitrary measure on $\Sigma = \mathcal{B}(\mathbb{R})$

Set
$$\mu(E) = \begin{cases} 1 & \text{if } 0 \in E \\ 0 & \text{else.} \end{cases}$$

Check $\mu(E)$ satisfies the properties of a measure.
This is called an "atomic" or "Dirac" measure.

Plan 3.2: First define "outer measure" on $\mathcal{P}(\mathbb{R}^n)$

outer measure on $\mathcal{P}(\mathbb{R}^n)$ $\xrightarrow{\text{when defined.}}$ Lebesgue measure on $\mathcal{M}(\mathbb{R}^n)$

Def 3.3: Let $E \subseteq \mathbb{R}^n$ be a subset. Let

$$m^*(E) = \inf \left\{ \sum_{j=1}^{\infty} |Q_j| \mid E \subseteq \bigcup_{j=1}^{\infty} Q_j \text{ is a covering by cubes} \right\}$$

Ex 3.4: $m^*(pt) = 0$ since $pt = x_0$, $Q = \prod [x_0^i - \frac{\varepsilon}{2}, x_0^i + \frac{\varepsilon}{2}]$ has volume $\varepsilon > 0$ for arbitrary ε .

Ex 3.5: $m^*(Q) = |Q|$ for a cube Q . Obvious Q covers itself so $m^*(Q) \leq |Q|$. Must show $|Q| \leq \sum_{j=1}^{\infty} |Q_j|$ for $Q \subseteq \bigcup Q_j$.

For each j , let $Q_j \subseteq S_j$ open w/ $|S_j| \leq (1+\varepsilon)|Q_j|$ for $\varepsilon > 0$.

$Q_j \subseteq \bigcup_{i=1}^N S_j$ by compactness,

$$|Q| \leq \sum_{j=1}^{\infty} |S_j| \leq (1+\varepsilon) \sum_{j=1}^{\infty} |Q_j| \leq (1+\varepsilon) \sum_{n=1}^{\infty} |Q_n|, \text{ take } \varepsilon \rightarrow 0$$

Ex 3.6 : Q an open cube, $|Q| = m^*(Q)$ still holds. (3.2)

Q covers, so $m^*(Q) \leq |Q|$. There is a closed $Q_0 \subseteq Q$ with volume $|Q_0| \geq (1-\varepsilon)|Q| \forall \varepsilon$, and

$$(1-\varepsilon)|Q| \leq |Q_0| \leq m(Q)$$

since any covering of Q covers Q_0 . Now take $\varepsilon \rightarrow 0$.

Ex 3.7 : $m^*(\mathbb{R}^n) = \infty$, obviously.

3.ii) Properties of the Outer Measure

Prop 3.8 (Monotonicity) If $E_1 \subseteq E_2$, then $m^*(E_1) \leq m^*(E_2)$.

Proof: Coverings of E_2 also cover E_1 .

Prop 3.9 (countable sub-additivity) If $E = \bigcup_{j=1}^{\infty} E_j$, then $m^*(E) \leq \sum_{j=1}^{\infty} m^*(E_j)$.

Proof: Can show for $m^*(E_j) < \infty$, else holds trivially.

Let Q_k^j be a collection of cubes covering E_j with

$$\begin{aligned} \sum_{k=1}^{\infty} |Q_k^j| &\leq m^*(E_j) + \frac{\varepsilon}{2^j} \\ m^*(E) &\leq \sum_{j,k=1}^{\infty} |Q_k^j| \leq \sum_{j=1}^{\infty} \sum_{k=1}^{\infty} |Q_k^j| \\ &\leq \sum_{j=1}^{\infty} m^*(E_j) + \varepsilon \quad \text{take lim } \varepsilon \rightarrow 0 \quad \square. \end{aligned}$$

Prop 3.10: $m^*(E) = \inf_{\substack{U \subseteq \mathbb{R}^n \\ \text{open } E \subseteq U}} m^*(U)$.

Proof: $m^*(E) \leq m^*(U) \forall E \subseteq U$ by Prop 3.8. Let Q_j be a covering so

$$\sum |Q_j| \leq m^*(E) + \varepsilon/2.$$

Take $U_j =$ open cube containing Q_j w/ area $\leq |Q_j| + \frac{\varepsilon}{2^{j+1}}$.

$$m^*(\bigcup U_j) \leq \sum m^*(U_j) \leq \sum |Q_j| + \frac{\varepsilon}{2^{j+1}} \leq \sum |Q_j| + \frac{\varepsilon}{2} \leq m^*(E) + \varepsilon \quad \square$$

Prop 3.11 (Translation Invariance)

(3.3)

Suppose A is a set and $A+y = \{u+y \mid u \in A\}$ for a fixed $y \in \mathbb{R}^n$.

Then $m^*(A) = m^*(A+y)$.

Proof: If Q is a cube, then $Q+y$ is also, so a covering Q_j of A gives $m^*(A+y) \leq \sum_{j=1}^{\infty} |Q_j+y| = \sum_{j=1}^{\infty} |Q_j| \leq m^*(A) + \varepsilon$.

For reverse, note $(A+y)-y = A$. □

Corollary 3.12: If $E \subseteq \mathbb{R}^n$ is countable, $m^*(E) = 0$.

Proof: $E = \bigcup_{j=1}^{\infty} E_j$ where $E_j = \text{pt.}$ Follows from Prop 3.9.

3.iii) Measurable Sets

Def 3.13: A subset $E \subseteq \mathbb{R}^n$ is said to be measurable if for any $A \subseteq \mathbb{R}^n$,

$$m^*(A) = m^*(A \cap E) + m^*(A \cap E^c).$$

ie E cuts any A into two pieces. (Note $\leq$ holds by subadditivity)

Ex 3.14: If $I \subseteq \mathbb{R}$ is an open interval, I is measurable.

Proof: Let $A \subseteq \mathbb{R}$ and let $A \subseteq \bigcup_{j=1}^{\infty} Q_j$ be a covering

Let $Q_j' = I \cap Q_j \Rightarrow Q_j'$ cover $A \cap I$

$Q_j'' = I^c \cap Q_j \Rightarrow Q_j''$ cover $A \cap I^c$

$$\begin{aligned} m^*(I \cap A) + m^*(I^c \cap A) &\leq \sum_{j=1}^{\infty} |Q_j'| + \sum_{j=1}^{\infty} |Q_j''| + \varepsilon \\ &= \sum_{j=1}^{\infty} |Q_j| + \varepsilon \leq m^*(A) + 2\varepsilon. \end{aligned}$$

and now take $\lim \varepsilon \rightarrow 0$.

Ex 3.15: Same for closed intervals.

Ex 3.15.5: Same for $\mathbb{R}^n$ but subdivide



Prop 3.16 : If E_j are disjoint, measurable, then $\bigcup_{j=1}^{\infty} E_j$ is measurable and

$$m^*(A \cap \bigcup_{j=1}^{\infty} E_j) = \sum m^*(A \cap E_j).$$

3.4

Proof: If $N=1$, this is vacuous. Take $A \in \mathbb{R}^n$ then set $A' = A \cap (E_1 \cup E_2)$

$$\begin{aligned} m^+(A') &= m^+(A' \cap E_1) + m^+(A' \cap E_1^c) \quad (E_1 \text{ measurable}) \\ &= m^+(A \cap E_1) + m^+(A \cap E_2) \end{aligned}$$

now use induction.

Thm 3.17 : The measurable sets form a σ -algebra, and $m^* : \mathcal{M}(\mathbb{R}^n) \rightarrow [0, \infty]$ is a measure.

Proof : i) $\mathbb{R}^n$ is measurable, since $m^*(A) = m^*(A \cap \mathbb{R}^n) + \underline{0}$
 ii) E measurable $\Rightarrow E^c$ measurable by symmetry of def. $\bar{m}^*(A \cap \emptyset)$.
 iii) Suppose E_i are measurable, then wts $\bigcup_{i=1}^{\infty} E_i$ is.

Step 1: True for finite unions/intersections. Suppose E, F measurable

$$\begin{aligned} m^*(A) &= m^*(A \cap E) + m^*(A \cap E^c) \\ &= m^*(A \cap E \cap F) + m^*(A \cap \underline{E \cap F^c}) + m^*(A \cap \underline{E^c \cap F}) \\ &= m^*(A \cap E \cap F) + m^*(A \cap (E \cap F)^c) \quad \text{by Prop 16.} \end{aligned}$$

Step 2 : Suppose E_i measurable, then

$$E'_i = E_i - \bigcup_{j=1}^{i-1} E_j = E_i \cap \left(\bigcup_{j=1}^{i-1} E_j \right)^c$$

is measurable, and $\bigcup E'_i = \bigcup E_i$, so suffices to assume disjoint. Now let $F_N = \bigcup_{i=1}^N E_i$ (measurable)

$A \in \mathbb{R}^n$ arbitrary.

$$\begin{aligned} m^*(A) &= m^*(A \cap \bigcup_{i=1}^{\infty} E_i) + m^*(A \cap F_n^c) \\ &= \sum_{i=1}^{\infty} m^*(E_i \cap A) + m^*(A \cap F_n^c) \\ &\geq \sum_{i=1}^{\infty} m^*(A \cap E_i) + m^*(A \cap E^c) \quad \text{by monotonicity} \end{aligned}$$

Take lim

$$\begin{aligned} &\geq \sum_{i=1}^{\infty} m^*(E_i \cap A) + m^*(A \cap E^c) \\ &\geq m^*(E \cap A) + m^*(E^c \cap A) \text{ by countable subadditivity.} \end{aligned}$$

and $\mu(E) \geq 0$, $\mu(\emptyset) = 0$. If $E = \bigcup E_j$ disjoint,

3.5

$$\mu(E) \leq \sum_{j=1}^{\infty} m^*(E_j) \quad (\text{additivity}).$$

$$\sum_{j=1}^{\infty} m^*(E_j) \leq \sum_{j=1}^{\infty} \sum_{k=1}^{\infty} |Q_{j,k}| \quad \text{covering}$$

$\therefore$ take $A = E$ in step 2.

Def 3.18 : $\mathcal{M}(\mathbb{R}^n)$ are called the measurable sets,
and $m^*|_{\mathcal{M}(\mathbb{R}^n)}$ is called the Lebesgue measure.

Lecture 4 Properties of Lebesgue Measure + Counterexamples.

[4.1]

Recall last time we showed $(\mathbb{R}^n, \mathcal{M}(\mathbb{R}^n), m)$ is a measure space.

Thm 2.15 : $(\mathbb{R}^n, \mathcal{M}(\mathbb{R}^n), m)$ is a measure space, and additionally satisfies

(1) Normalization $m(Q) = |Q|$

(2) Complete $E \in \mathcal{M}(\mathbb{R}^n)$ then $m(E) = 0 \Rightarrow A \subseteq E$ measurable

(3) translation invariant $E \in \mathcal{M}(\mathbb{R}^n) \Rightarrow E+y \in \mathcal{M}(\mathbb{R}^n)$ w/ $m(A) = 0$.
w/ $m(E) = m(E+y)$

(4) Borel complete $\mathcal{B}(\mathbb{R}^n) \subseteq \mathcal{M}(\mathbb{R}^n)$.

Proof : (1) For open/closed Q , $m(Q) = m^*(Q) = |Q|$.

(2) Suppose E has ^{outer} measure 0. Then $A \cap E \subseteq E$, $A \cap E^c \subseteq A$
so by monotonicity

$$m^*(A) \geq m^*(A \cap E^c)$$

$$\geq m^*(A \cap E^c) + m^*(A \cap E) \stackrel{0}{=} m^*(A) \Rightarrow E \text{ is measurable.}$$

If $A \subseteq E$, then $m^*(A) = 0$ so also measurable.

(3) $A \cap (E+x) = (A-x) \cap E$ ~~so~~, and m^* is translation inv.

(4) open sets $\in \mathcal{M}(\mathbb{R}^n)$ and it's a σ -algebra, $\square$

4.i) Continuity of Measure

Prop 4.1 : Suppose $E = \bigcap_{i=1}^{\infty} E_i$ with $E_1 \supseteq E_2 \supseteq \dots$ and $m(E_1) < \infty$.

$$\text{Then } m(E) = \lim_{N \rightarrow \infty} m(E_N).$$

Proof : $E_1 = E \cup (E_1 \setminus E_2) \cup (E_2 \setminus E_3) \cup \dots$

$$m(E_1) = m(E) + \sum_{n=1}^{\infty} m(E_n \setminus E_{n+1}) = m(E) + m(E_1) - \lim_{i \rightarrow \infty} m(E_i)$$

$$\Rightarrow m(E) = \lim m(E_i) \quad \square$$

Rem 4.2 : Informally we can write

$$m(\lim E_i) = \lim m(E_i)$$

which looks more like standard continuity.

Lemma 4.2 (Borel - Cantelli) Suppose E_i measurable, $\sum_{n=1}^{\infty} m(E_n) < \infty$. [4.2]

Then $m\{x \mid x \in E_n \text{ for } \infty\text{-many } n\} = 0$.

Proof: Call this set A . $A \subseteq \bigcup_{n=N}^{\infty} E_n$ for any N .

$$m(A) \leq m\left(\bigcup_{n=N}^{\infty} E_n\right) \leq \sum_{n=N}^{\infty} m(E_n) \rightarrow 0$$

by convergence.

4.ii) Some funky sets.

Ex 4.3 (Cantor Set) $K \subseteq [0,1]$ a set uncountable but $m(K)=0$.

$$K_0 = [0,1]$$

$$K_1 = K_0 \setminus (\text{middle third})$$

$$K_2 = K_1 \setminus (\text{middle thirds})$$



⋮

$$K = \bigcap_{n=0}^{\infty} K_n$$

$$\text{Clearly } m(K_n) = \frac{2}{3} m(K_{n-1}) = \left(\frac{2}{3}\right)^n$$

$$m(K) = \lim_{n \rightarrow \infty} m(K_n) = \lim_{n \rightarrow \infty} \left(\frac{2}{3}\right)^n = 0.$$

Claim: K is uncountable. $K = \{x \in [0,1] \mid x = 0.x_1x_2x_3 \dots \text{ 'in base 3'}$
(bijective w/ $\mathbb{R}$) $\wedge x_i = 0, 2$.

Suppose we can enumerate $K_1, K_2, \dots \Rightarrow$ Cantor's diagonalization $\square$.

Corollary 4.4: There exist measurable sets that are not Borel.
In fact, "almost all" are such.

Proof: $m(K)=0$, so any $A \subseteq K$ is measurable with measure 0.

$\Rightarrow \mathcal{P}(K) \subseteq \mathcal{M}(\mathbb{R})$, but $\mathcal{P}(K)$ has cardinality $2^{\aleph_0} \geq$
Borel sets have cardinality $\aleph_1$.

Ex 4.5 (Normal Numbers)

[4.3]

Def 4.6: $x \in [0,1]$ is normal if any finite sequence of digits occurs infinitely often.

Lemma 4.7: $m\{x \in [0,1] \mid x \text{ normal}\} = 1$.

Proof: Let $E_i = \{x \in [0,1] \mid i^{\text{th}} \text{ digit is } 1\}$. Clearly $m(E_i) = \frac{1}{10}$.

$A = \{x \mid 1 \text{ occurs inf often}\}$

$$A^c = \{x \mid \exists N \text{ st } x \notin \bigcup_{n=N}^{\infty} E_n\} = \bigcup_{N=1}^{\infty} \bigcap_{n=N}^{\infty} E_n^c$$

$$m(A^c) = m\left(\bigcup_{N=1}^{\infty} \bigcap_{n=N}^{\infty} E_n^c\right) \leq \sum_{N=1}^{\infty} m\left(\bigcap_{n=N}^{\infty} E_n^c\right) = 0.$$

= 0 by Borel-Cantelli.

$$\Rightarrow m(A) = 1.$$

$E_i' = \{x \in [0,1] \mid i^{\text{th}} - i \cdot 10^{\text{th}} \dots\}$ same proof.

Conjecture 4.8: $\pi, \sqrt{2} - 1$ are normal.

Corollary 4.9: It is not known if π contains Hamlet in some decimal form.

Ex 4.10: There exists a non-measurable set.

Let $N = \{ \text{a lift of } \mathbb{R} \downarrow \mathbb{R}/\mathbb{Q} \text{ ie a choice of representatives of cosets } x \sim y \text{ if } x - y \in \mathbb{Q} \}$.

this invokes the axiom of choice!

Note each $x \in \mathbb{R}$ can be written $x = a + q$ uniquely w/ $a \in N, q \in \mathbb{Q}$.

Lemma 4.11: N is not measurable.

Proof: Suppose it is. Set $S = [0,1] \cap N$.

$$B_{m,s} = \{s + a \mid s \in S, a \in [m, m+1]\}$$

$\bigcup_{s \in S} B_{m,s} \subseteq [m, m+2]$, and all disjoint, and translates, ~~so $\sum m(B_{m,s})$~~

Therefore

$$\sum_{s \in S} m(B_{m,s}) \leq 2 \Rightarrow m(B_{m,s}) = 0 \text{ b/c all equal.}$$

In particular, $m(N \cap [m, m+1]) = m(B_{m,0}) = 0 \Rightarrow m(N) = 0$.

But

$$\infty = m(\mathbb{R}) = m\left(\bigcup_{q \in \mathbb{Q}} N+q\right) = \sum 0 = 0 \quad \rightarrow \leftarrow \quad \square$$

Corollary 4.12 : If $E \subseteq \mathbb{R}$ has positive measure, $\exists$ a $N \subseteq E$ non-measurable.

Remark 4.13

Thm 4.13 (Solovay 1970) : A non-measurable ^{Lebesgue} set cannot be constructed without the Axiom of Choice.

Rem 4.14 : If A is measurable $A - A = \{x - y \mid x, y \in A\}$ need not be measurable!

Note also $K - K = [-1, 1]$ despite $m(K) = 0$.

Open Problem 4.15 : If $x \in K$, then can x be algebraic?
and $x \in \mathbb{Q}$

Lecture 5 Measurable Functions

[5.1]

Def 5.1 : A property P is said to hold "almost everywhere" if it holds on a set of full measure.

5.1) Littlewood's First Principle

Principle 5.2 : "Every measurable set is closed to a finite union of intervals"

Def 5.3 : Given two set A, B the symmetric difference
 $A \Delta B = (A \setminus B) \cup (B \setminus A)$

Prop 5.4 : If $m(E) < \infty$, then $\forall \varepsilon > 0$,
 $\exists J$ a finite union of intervals st
 $m(J \Delta E) < \varepsilon$.



Proof : Note that $m(B \setminus A) = m(B) - m(A)$
since $m(B) = m(B \cap A) + m(B \cap A^c)$
 $= m(A) + m(B \setminus A)$

Take $E \subseteq \bigcup Q_n$ such that
 $\sum |Q_n| \leq m(E) + \varepsilon/2$.

Now take N large so that

$$\sum_{n=N+1}^{\infty} |Q_n| < \frac{\varepsilon}{2}$$

$$m(E \setminus \bigcup_{n=1}^N Q_n) \leq m(\bigcup_{n=1}^{\infty} Q_n \setminus \bigcup_{n=1}^N Q_n) < \varepsilon/2$$

$$m(\bigcup_{n=1}^N Q_n \setminus E) \leq m(\bigcup_{n=1}^{\infty} Q_n \setminus E) < \varepsilon/2 \quad \square$$

Def 5.5 : A G_δ set is a countable intersection of opens ^[5.2]
an F_σ set is a countable union of closed

Note both are Borel.

Prop 5.6 : A set $E \subseteq \mathbb{R}^n$ is measurable : $\text{iff } \exists$

$$F_\sigma \subseteq E \subseteq G_\delta$$

such that $m(E \setminus F) = m(G \setminus E) = 0$.

Proof : If the statement holds, E is measurable
since F, G are and all measure 0 sets are.

Let U_n be open with $m(U_n \setminus E) \leq \frac{1}{n}$,
and set $G_\delta = \bigcap U_n$, and $m(G \setminus E) \leq \frac{1}{n} \forall n$.

Let V_n be open w/ $m(V_n \setminus E^c) \leq \frac{1}{n}$.

$$F_\sigma = \bigcup V_n^c$$

suffices.

5.ii) Measurable Functions (We will work only on $\mathbb{R}$)
for now

Prop 5.7 : TFAE for $f: \mathbb{R} \rightarrow \mathbb{R}$

1) $f^{-1}(a, \infty]$ is measurable $\forall a \in \mathbb{R}$.

2) $f^{-1}(U)$ is measurable $\forall U \subseteq \mathbb{R}$ open

3) $f^{-1}(B)$ is measurable $\forall B \subseteq \mathbb{R}$ Borel.

Proof : $3 \Rightarrow 2 \Rightarrow 1$ a fortiori.

Let $\Sigma(f) = \{ A \in \mathcal{P}(\mathbb{R}) \mid f^{-1}(A) \text{ is measurable} \}$

It is easy to check $\Sigma(f)$ is a σ -algebra, and (a, ∞)
generates $\mathcal{B}$ so $\mathcal{B}(\mathbb{R}) \subseteq \Sigma(f)$

Prop 5.8: $\mathcal{M}(\mathbb{R})$ is an algebra containing $\mathcal{C}^0(\mathbb{R})$.

Proof: Obviously $\mathcal{C}^0 \subseteq \mathcal{M}$ by 2) of Prop 5.7.

1) $f \in \mathcal{M}(\mathbb{R})$ then $\alpha f \in \mathcal{M}(\mathbb{R})$ for $\alpha \in \mathbb{R}$ is clear, because $f^{-1}(a, b) = \alpha f^{-1}(\frac{a}{\alpha}, \frac{b}{\alpha})$.

2) Suppose $f, g \in \mathcal{M}(\mathbb{R})$. Suppose $x \in (f+g)^{-1}(a, \infty)$.

Let $\frac{2}{q} > a - g(x)$ be a rational, such a $\frac{2}{q}$ exists iff $f(x) + g(x) > a$.

$$(f+g)^{-1}(a, \infty) = \bigcup_{\substack{p/q \in \mathbb{Q} \\ p/q < f(x) + g(x)}} \{x \mid f(x) > p/q\} \cap \{x \mid \frac{2}{q} + g(x) > a\}$$

$\in \mathcal{M}(\mathbb{R})$.

↑ countable union of measurable.

3) If $f, g \in \mathcal{M}(\mathbb{R})$ then $f \cdot g \in \mathcal{M}(\mathbb{R})$.

$\frac{1}{2}[(f+g)^2 - f^2 - g^2] = f \cdot g$ so by 1), 2) enough to show for

$f = g$. $(f^2)^{-1}(a, \infty) = f^{-1}(\sqrt{a}, \infty) \cup f^{-1}(-\infty, -\sqrt{a})$. $\square$

Warning 5.9: $\mathcal{M}(\mathbb{R})$ is not closed under composition!

but $\mathcal{C}^0(\mathbb{R}) \times \mathcal{M}(\mathbb{R}) \rightarrow \mathcal{M}(\mathbb{R})$

$(h, f) \rightarrow h \circ f$ is measurable.

Theorem 5.10: Suppose f_n is a sequence of measurable functions such that $f_n \rightarrow f$ pointwise, then $f \in \mathcal{M}(\mathbb{R})$.

Proof: For $c \in \mathbb{R}$, write

$$\begin{aligned} f^{-1}(c, \infty) &= \{x \mid \exists k \in \mathbb{N} \text{ st } \exists N \in \mathbb{N} \text{ st } \forall n \geq N, f_n(x) > c + \frac{1}{k}\} \\ &= \bigcup_{k \in \mathbb{N}} \bigcup_{N \in \mathbb{N}} \bigcap_{n \geq N} \{x \mid f_n(x) > c + \frac{1}{k}\}. \end{aligned}$$

Countable $\cup$ and $\cap$ of measurable

$\square$.

Def 5.11 : The Indicator function of a measurable set A is

$$\chi_A = \begin{cases} 1 & x \in A \\ 0 & \text{else} \end{cases}$$

It is clearly measurable.

Def 5.12 : A function is called simple if it is a finite sum

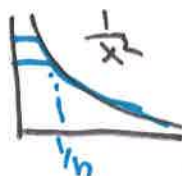
$$f = \sum_{i=1}^N \alpha_i \chi_{E_i} \quad E_i \text{ measurable.}$$

and a step function if E_i are disjoint intervals.

Def 5.13 : f_n converges to f in measure if $\forall \varepsilon > 0$.

$$\lim_{n \rightarrow \infty} m\{x \mid |f_n(x) - f(x)| > \varepsilon\} = 0.$$

Ex 5.14 :



$$f_n = \min(n^2, \frac{1}{x^2})$$

then $f_n \rightarrow f$ in measure

but $\sup |f_n - f| = \infty$ so not uniformly

Thm 5.15 Suppose $f: [a, b] \rightarrow \mathbb{R}$ is measurable. $\exists f_n \rightarrow f$ in measure such that f_n are

- i) bounded and measurable
- ii) Simple
- iii) Step
- iv) Continuous.

Proof i) let $f_n = \begin{cases} f(x) & |f(x)| \leq n \\ 0 & \text{else} \end{cases}$

then $|f_n - f| = 0$ outside $E_n = \{x \mid |f(x)| > n\}$.

Since $\bigcap E_n = \emptyset$ and $m(E_n) < \infty$, $m(E_n) \rightarrow 0$ by continuity of measure.

ii) By diagonalization can assume f bounded. Say $|f(x)| \leq M$.

$$\text{Let } I_{nk} = [-M + \frac{k}{n}, -M + \frac{(k+1)}{n}]$$

$$E_k = \{x \mid f(x) \in I_{nk}\}$$

Take $f_n = \sum_{k=1}^{2Mn} (-M + \frac{k}{n}) \chi_{E_k}$ $|f_n - f| \leq \frac{1}{n}$. Now send $n \rightarrow \infty$.

iii) Approximating χ_{E_k} by step is same as approximating E by opens.



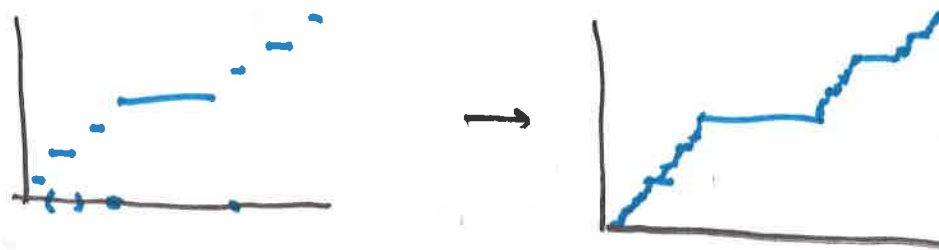
Lecture 6 Measurable Functions II: Littlewood's 2nd - 3rd Principles.

6.1

6.i) The Cantor Function / Devil's Staircase

Ex 6.1: Let $c(x): [0,1] \rightarrow [0,1]$ be defined by

$$c(x) = \begin{cases} \frac{1}{2} & x \in (\frac{1}{3}, \frac{2}{3}) \\ \frac{1}{4}, \frac{3}{4} & x \in (\frac{1}{9}, \frac{2}{9}) \text{ or } (\frac{7}{9}, \frac{8}{9}) \end{cases} \quad \text{on } K^c = \sup \{c(t) \mid t \in K_n^c \cap [0,x]\} \text{ on } K$$



Lemma 6.2: $C(x): [0,1] \rightarrow [0,1]$ is continuous (hence measurable) and monotonically ^{non-dec.} increasing, but $C'(x) = 0$ a.e.

Proof: Clearly continuous on K^c . If $x \in K$, then x lies between two intervals of K_n^c w/ $K^c = \bigcup_n K_n^c$, for ^{any} n , and these have $|C(x) - C(y)| \leq \frac{1}{2^n}$.
Monotonically non-decreasing is obvious from sup.
 $C'(x)$ exists and $= 0$ on K^c and $m(K^c) = 1$.

Corollary 6.3: (Even though it's not precisely defined yet)

$$\int_0^1 C'(x) dx = 0 \cdot m(K^c) + ? \cdot m(K) = 0$$

So in particular, the FTC fails for C ,

$$0 = \int_0^1 C'(x) dx \neq C(1) - C(0) = 1.$$

Corollary 6.4: $\exists \varphi$ a homeomorphism $\varphi: [0,1] \rightarrow [0,2]$ that sends a set of measure 0 to a set of positive measure, and vice-versa.

Proof: $\varphi(x) = x + C(x)$ is ^(strictly) monotonically increasing so a homeomorphism $K \rightarrow \text{measure } 1$.
 $\frac{1}{2}\varphi^{-1}(2x)$ has opposite property.

6:ii) Littlewood's principles II

6.2

Principle 6.5 (Littlewood I: Measurable Sets): Any measurable set is nearly a finite union of open intervals

Principle 6.6 (Littlewood II: Measurable Functions): Any measurable function is nearly continuous

Principle 6.7 (Littlewood III: Convergence): Any convergent sequence is nearly uniformly convergent.

Here's a formal version of II:

Thm 6.8 (Lusin) Let $f: \mathbb{R} \rightarrow \mathbb{R}$ be measurable. Then given $\varepsilon > 0$, $\exists g: \mathbb{R} \rightarrow \mathbb{R}$ continuous st. $f \equiv g$ outside a set of measure $< \varepsilon$.

Recall we can construct $f_n \rightarrow f$ where $f: [a, b] \rightarrow \mathbb{R}$
 f_n bounded, simple, step, $\mathcal{C}^0$.

Proof: 1) It suffices to consider $f: [n, n+1] \rightarrow \mathbb{R}$, then use the " $\frac{\varepsilon}{2^n}$ " trick.

2) Since $f_n \rightarrow f$ in measure, f_n bounded, can assume f bounded.

3) $g_n \in \mathcal{C}^0$ st $g_n \rightarrow f$ in measure
 $\Rightarrow \exists g_1$ st $m(\underbrace{|g_1 - f|}_{h_1} > \frac{1}{2}) < \frac{\varepsilon}{2}$.

$$f = g_1 + h_1$$

Now $\exists g_2$ st $m(|g_2 - g_1| > \frac{1}{2^2}) < \frac{\varepsilon}{4}$
Inductively

$$f = g_1 + \dots + g_N + h_N$$

g_i continuous, $|h_N| \leq \frac{1}{2^N}$ except on $E = \bigcup_{j=1}^N E_j$
with $m(E_j) \leq \frac{\varepsilon}{2^j}$, $h_N = 0$ on E^c .

4) Let $\sup |f_N(x)| \leq \sup |h_{N-1}| \leq \frac{1}{2^{N-1}}$, $g = \sum g_i$ converges uniformly
and $f = g$ outside $m(\bigcup E_j) = \varepsilon$.

6.iii) Convergence

6.3

There are three types of convergence discussed so far.

$$\left\{ \begin{array}{l} f_n \rightarrow f \\ \text{in measure} \end{array} \right\} \overset{\text{implies}}{\subseteq} \left\{ \begin{array}{l} f_n \rightarrow f \\ \text{pointwise} \\ \text{(a.e.)} \end{array} \right\} \overset{\text{implies}}{\subseteq} \left\{ \begin{array}{l} f_n \rightarrow f \\ \text{uniformly} \end{array} \right\}$$

$m(|f_n - f| > \varepsilon) \rightarrow 0 \quad \forall \varepsilon$
 $f_n(x) \rightarrow f \quad \forall x$
 $\sup_x |f_n(x) - f(x)| \rightarrow 0$
 $= \text{Convergence in } \ell^\infty.$

Prop 6.9: Suppose $f_n \rightarrow f$ ~~in measure~~ pointwise a.e. on $[a, b]$

then $f_n \rightarrow f$ in measure

Proof: Let $\varepsilon > 0$. $E_n = \{|f_n - f| > \varepsilon\}$ ~~$\bigcap_{n=1}^{\infty} E_n = \emptyset$ by pointwise a.e.~~

~~$\lim_{n \rightarrow \infty} m(E_n) = m(\bigcap_{n=1}^{\infty} E_n) = 0$~~ so

$B_N = \bigcup_{n=N}^{\infty} E_n$ so $B_N \supseteq B_{N+1} \supseteq \dots$

$\lim_{n \rightarrow \infty} m(E_n) \leq \lim_{N \rightarrow \infty} m(B_N) \overset{\text{cont. of measure}}{=} m\left(\bigcap_{N=1}^{\infty} B_N\right)$

$= m\left\{x \mid \exists n \geq N \text{ w/ } |f_n - f| > \varepsilon\right\}$

$= 0 \quad \text{by pointwise a.e.}$

Ex 6.10: Note convergence $f_n \rightarrow f$ in measure for f_n bounded cannot happen on $\mathbb{R}$ (on any inf. measur. set)

$f(x) = x$ is a counter example, However, any such f is still a lim in measure of f cont.

Ex 6.11: Prop 6.9 also fails on $\mathbb{R}$. b/c $m(B_1) = \infty$ is possible

$f_n = \sin x \quad \forall [-n, n] \rightarrow \sin x$ pointwise but

$m(|f_n - f| > \frac{1}{2}) = \infty \quad \forall n.$

Ex 6.11 : $\exists f_n \rightarrow f$ in measure but not pointwise.

Let I_n = interval of length $\frac{1}{\sqrt{n}}$.

"limp function"



so $\bigcup_{n=1}^{\infty} I_n = [0,1] \forall N$. so $\sum |I_n| = \infty$.

Then $f_n = \chi_{I_n} \rightarrow 0$ in measure, but $f_n \rightarrow 0$ pointwise nowhere.

Thm 6.12 : $f_n \rightarrow f$ in measure on $\mathbb{R}$, $\exists$ a subseq converging pointwise a.e.

Proof : $E_n = \{|f_n - f| > \frac{1}{2^n}\}$, take n_k such that $\{m(E_{n_k}) < \frac{1}{k^2}\}$. Then $f_n \rightarrow f$ outside $\bigcup_{k=1}^{\infty} E_{n_k}$ but this $\rightarrow 0$ w/k because $\sum m(E_{n_k}) < \infty$.

6.14 : Littlewood's 3rd Principle

Precise version:

Thm 6.13 (Egoroff): Let $f_n \rightarrow f$ pointwise a.e. on $[a,b]$.

$\forall \varepsilon > 0$, $\exists E_{\varepsilon} \subseteq [a,b]$ st. $m([a,b] \setminus E_{\varepsilon}) < \varepsilon$ and $f_n \rightarrow f$ uniformly on E_{ε} .

Proof : Set $\delta_n = \sup_{i \geq n} |f_i(x) - f(x)|$

Note $\delta_n \geq \delta_{n+1}$ and $\delta_n \rightarrow 0$ pointwise a.e.

$\exists n_k$ st. $\delta_{n_k} \rightarrow 0$ in measure, so $\forall k \exists$

E_k st. $m(E_k) \leq \frac{\varepsilon}{2^k}$ and $|\delta_{n_k}| \leq \frac{1}{k}$ on E_k^c .

Then $\delta_{n_k} \rightarrow 0$ uniformly on $E_{\varepsilon}^c \cap E_k^c$ and

$$m(E_{\varepsilon}) \leq m([a,b]) - \sum E_k \geq (b-a) - \sum \frac{\varepsilon}{2^k} \geq b-a-\varepsilon.$$

But δ_{n_k} decreasing so conv on subseq $\Rightarrow$ conv of whole seq $\square$

Lecture 7 Lebesgue Integration I: bounded functions

(7.1)

Assumption 7.1 : Throughout this lecture, assume E measurable w/ $m(E) < \infty$ and all measurable functions lie in $\mathcal{M}_B(E) = \{ f \in \mathcal{M}(E) \mid \exists M \text{ w/ } |f| \leq M \}$.

Def 7.2 : Define the (Lebesgue) Integral of a function $f \in \mathcal{M}_B(E)$

w/ $m(E) < \infty$, denote

$$\int_E dx : \mathcal{M}_B(E) \rightarrow \overline{\mathbb{R}}$$

$$\text{or } \int_E dm(x) \quad \begin{array}{c} \uparrow \\ \text{Lebesgue} \\ \text{measure} \end{array}$$

By

$$\int_E f dx := \sup_{\substack{\varphi \text{ simple} \\ \varphi \leq f}} \left\{ \sum_{i=1}^N a_i m(E_i) \mid \varphi = \sum_{i=1}^N a_i \chi_{E_i} \right\}$$

Def 7.3 : $f \in \mathcal{M}_B(E)$ is said to be (absolutely) Lebesgue integrable if $\int_E |f| dx < \infty$

Rem 7.4 : If $E' \subseteq E$, then one can extend

$$- : \mathcal{M}_B(E') \rightarrow \mathcal{M}_B(E)$$

and

$$f \mapsto \bar{f} = \begin{cases} f & x \in E' \\ 0 & \text{else} \end{cases}$$
$$\int_{E'} f dx = \int_E \bar{f} dx.$$

Prop 7.5 : $\int_E dx : \mathcal{M}_B(E) \rightarrow \overline{\mathbb{R}}$ satisfies

1) $\int_E af + bg dx = a \int_E f dx + b \int_E g dx$ (linearity)

2) $\int_E \chi_A dx = m(A)$ for $A \subseteq E$ measurable (normalization)

3) $f \leq g \Rightarrow \int_E f dx \leq \int_E g dx$ (monotonicity)

4) $\left| \int_E f(x) dx \right| \leq \int_E |f(x)| dx$

5) $\int_E |f(x)| dx \leq m(E) \sup |f(x)| < \infty.$

Rem 7.6: A simple function $\varphi: E \rightarrow \mathbb{R}$ has a canonical representation ^(7.2)

$$\varphi = \sum_{k=1}^N \lambda_k \chi_{E_k}$$

where $E_k = \varphi^{-1}(\lambda_k)$ for all k . Other representations arise from subdividing or repetition. We assume all simple functions are in this canonical representation. (See Royden 4.2.1)

Prop 7.7: If $f \in \mathcal{M}_B(E)$ then

Note

$\int_E \varphi dx = \sum a_k m(E_k)$
is immediate

$$\sup_{\varphi \leq f} \int_E \varphi dx = \int_E f dx = \inf_{\varphi \geq f} \int_E \varphi dx$$

agree and are finite. In fact, if f bounded, this is iff.

Proof: $\Rightarrow$ For any choice $\varphi \leq f \leq \psi$, so suffices to show $|\inf - \sup| < \varepsilon$ for any ε . Given ε , let $|f| \leq M$, and divide $[-M, M] = \bigcup I_n = \bigcup [a_n, b_n]$ where $|I_n| < \varepsilon$.

Set $E_n = f^{-1}(I_n)$. Then

$$\sum_{n=1}^N a_n m(E_n) \leq \inf_{\varphi \leq f} \sum_{n=1}^N \alpha_k m(E_k) \quad \text{and on the other hand}$$

$$\inf_{\varphi \leq f} \sum_{n=1}^N \beta_k m(E_k) \leq \sum_{n=1}^N b_n m(E_n) \leq m(E) \cdot \varepsilon + \sum_{n=1}^N a_n m(E_n)$$

so $|\sup - \inf| \leq m(E) \cdot \varepsilon$.

$\Leftarrow$ Suppose $\sup = \inf$. Let φ_n, ψ_n be with $\frac{1}{2^n}$ of these so $\int_E |\varphi_n - \psi_n| dx \leq \frac{1}{n}$. monotonic, bounded above

Since $\varphi_n \leq f \leq \psi_n$, $\forall n$ $\lim_{n \rightarrow \infty} \varphi_n \leq f \leq \lim_{n \rightarrow \infty} \psi_n$

Thus $\varphi = \psi$ a.e. $\Rightarrow f = \varphi = \psi$ a.e. $\Rightarrow f$ measurable since φ, ψ are.

Claim: $\varphi_n \rightarrow \psi_n$ in measure. i.e. $\forall \varepsilon > 0$ $\lim_{n \rightarrow \infty} \frac{1}{\varepsilon} m(|\varphi_n - \psi_n| > \varepsilon) = 0$.

Pick $\delta > 0$. N st $\frac{1}{N} < \varepsilon \delta$. Then

$$m(|\varphi_n - \psi_n| > \varepsilon) \cdot \varepsilon \leq \int_E |\varphi_n - \psi_n| \leq \frac{1}{N} \leq \varepsilon \delta$$

$m(\quad) \leq \delta$ (now subseq conv. but monotone) $\square$

Proof (of Prop 7.5).

(7.3)

$$1) \int_E f dx + \int_E g dx \leq \sup_{\psi \leq f, \psi \leq g} \int \psi \leq \sup_{\psi \leq f+g} \int \psi dx = \int_E f+g.$$

Using inf get reverse.
 $\int_E \alpha f dx = \alpha \int_E f dx$ is obvious.

$$2) \text{ If } f = \chi_A \text{ then } \psi \leq \chi_A \Rightarrow \psi \leq 0 \text{ on } A^c, \psi \leq 1 \text{ on } A, \text{ so}$$
$$\int_E \psi dx \leq m(A), \text{ but } \psi = \chi_A \text{ itself gives } =.$$

$$3) \text{ If } f \leq g, \sup_{\psi \leq f} \leq \sup_{\psi \leq g} \text{ is the sup over a strictly larger set.}$$

$$4) -|f| \leq f \leq |f| \text{ so by 3, } -\int_E |f| dx \leq \int_E f dx \leq \int_E |f| dx$$

$$5) \text{ by inf def, } \inf \leq m(E) \cdot \sup |f|.$$

$$\sup \geq m(E) \cdot (-\sup |f|)$$

Prop 7.8 : If $A, B \in \mathcal{E}$ disjoint, measurable

$$\int_{A \cup B} f dx = \int_A f dx + \int_B f dx$$

Proof : By linearity

$$\begin{aligned} \int_{A \cup B} f dx &= \int_E f \chi_{A \cup B} dx = \int_E f (\chi_A + \chi_B) \\ &= \int_E f \chi_A dx + \int_E f \chi_B dx = \int_A f dx + \int_B f dx. \end{aligned}$$

Thm 7.9 (Bounded convergence) Suppose $f_n \in M_b(E)$ are uniformly 7.4
 bounded $|f_n| \leq M$, $f_n \rightarrow f$ pointwise a.e. Then

$$\lim_{n \rightarrow \infty} \int_E f_n dx = \int_E \lim f_n dx = \int_E f dx$$

Proof: By Littlewood III, $\exists A \subseteq E$ with $m(E \setminus A) < \frac{\varepsilon}{2M}$ st
 $\sup_{x \in A} |f_n(x) - f| \rightarrow 0$.

Now let $\varepsilon > 0$. n large enough so $\frac{\varepsilon}{2} \cdot m(E) \leq \frac{\varepsilon}{2} \cdot m(E)$.

$$\begin{aligned} \left| \int_E f dx - \int_E f_n dx \right| &\leq \left| \int_A f_n - f dx + \int_{A^c} f_n - f \right| \\ &\leq \int_A |f_n - f| dx + \int_{A^c} |f_n - f| dx \\ &\leq m(A) \sup_{x \in A} |f_n - f| + m(A^c) 2M \\ &\leq m(E) \cdot \frac{\varepsilon}{2} m(E) + \frac{\varepsilon}{4M} \cdot 2M \leq \varepsilon. \end{aligned}$$

7.ii) Riemann Integrals

Def 7.10: Recall $f: [a, b] \rightarrow \mathbb{R}$ is Riemann integrable if
 $\lim_{n \rightarrow \infty} \sum_{n=1}^N \frac{b-a}{n} \cdot \inf_{x \in I_n} f = \lim_{n \rightarrow \infty} \sum_{n=1}^{\infty} \frac{b-a}{n} \sup_{x \in I_n} f$

where $I_n = [a + \frac{b-a}{N}(n-1), a + \frac{b-a}{N} \cdot n]$.

Thm 7.11: If f is Riemann integrable, then it is Lebesgue integrable, and they agree.

Proof: $\frac{b-a}{n} \cdot \inf_{I_n} f = m(I_n) \cdot \inf_{I_n} f$ is a special case of a simple function,
 likewise for sup □

Lecture 8 Lebesgue Integration II: ~~functions~~ functions on $\mathbb{R}$. [8.1]

Throughout this lecture we assume $E = \mathbb{R}$ or has (possibly) infinite measure, and $f \in \mathcal{M}_+(E)$ is non-negative but not (necessarily) bounded (above).

8.i) Non-Negative functions ^{at first}

Def 8.2: The support of a function $f \in \mathcal{M}(\mathbb{R})$ is

$$\text{supp } \varphi = \{x \mid \varphi(x) \neq 0\}$$

which is measurable. It has finite support if $m(\text{supp } \varphi) < \infty$.

Recall if $f \in \mathcal{M}(E)$ then it may be extended by 0 $f'' = f \chi_E$.

Def 8.3: Define the Lebesgue Integral

$$\int_E dx : \mathcal{M}_+(E) \rightarrow \overline{\mathbb{R}}$$

by

$$\int_E f dx = \sup_h \left\{ \int_E h \mid h \in \mathcal{M}_B(E) \text{ has finite support, } 0 \leq h \leq f \right\}$$

Rem 8.4: By the way, $f \in \mathcal{M}(E)$ is tacitly allowed to take value $\pm\infty$ on a set of measure 0, thus e.g. $\frac{1}{x} \in \mathcal{M}(\mathbb{R})$ while technically perhaps we should write $\frac{1}{x} \in \mathcal{M}(\mathbb{R}, \{0\})$.

Prop 8.5: $\int : \mathcal{M}_+(E) \rightarrow \overline{\mathbb{R}}$ still satisfies

1) Linearity

2) $\int_E \chi_A dx = m(A)$

3) $f \leq g \Rightarrow \int_E f dx \leq \int_E g dx$

4) $\left| \int_E f(x) dx \right| \leq \int_E |f(x)|$

Proof: 3), 4) same as before. 2) same for $m(A) < \infty$, $m(A) = \infty$

1) $\int f + \int g \leq \int f + g$ as before

$\sup \rightarrow \infty$
is obvious.

$\geq$ used inf characterization, which we no longer have.

Let $h \leq f+g$ be bounded w/ finite support. Write

$$h = h_1 + h_2 \quad \text{w/ } h_1 = \min(h, f) \quad h_2 = h - h_1.$$

Note that $h_1 \leq f$ by construction

[8.2]

$h_2 \leq g$ (if $h_1 = f$, then $h_2 = 0 \leq g$
if $h_1 = f$, then $h_2 = h - h_1 \leq f + g - h_1 \leq g$)

Thus

$$\begin{aligned} \int_E h dx &= \int_E h_1 dx + \int_E h_2 dx && \text{(by linearity for bounded, finite supp)} \\ &\leq \int_E f dx + \int_E g dx \end{aligned}$$

$$\Rightarrow \sup \int_E h dx = \int_E f + g dx \leq \int_E f dx + \int_E g dx$$

$$\int_E \alpha f dx = \alpha \int_E f dx \text{ still immediate}$$

Prop 8.6: If A, B disjoint

$$\int_{A \cup B} f dx = \int_A f dx + \int_B f dx$$

continues to hold

Proof:

$$\begin{aligned} \int_{A \cup B} f dx &= \sup_{\substack{h \text{ bd} \\ f \leq h}} \int_{A \cup B} h = \sup_{\substack{h \text{ bd} \\ f \leq h}} \int_A h dx + \int_B h dx \\ &\leq \int_A f dx + \int_B f dx \end{aligned}$$

but also $\geq$ up to ε by taking $h = h_1 \chi_A + h_2 \chi_B$ for h_i within $\varepsilon/2$. $\square$

8.ii) General Functions: Now drop the assumption $f \geq 0$.

Def 8.7: Define $f \in \mathcal{M}(E)$ as Lebesgue integrable if

$$\int_E |f| dx < \infty.$$

Set $f^\pm = \max\{0, \pm f\}$, both measurable, and $\int_{\{f \geq 0\}} f^+ dx + \int_{\{f < 0\}} f^- dx = \int |f| dx < \infty$
so both are integrable.

Def 8.8: If $f \in \mathcal{M}(E)$ is Lebesgue integrable, define

$$\int_E f dx = \int_E f^+ dx - \int_E f^- dx.$$

Prop 8.7 : Properties (1)-(4) continue to hold.

2) is same since $\chi_A \geq 0$.

3) Follows from linearity because $f \leq g \Rightarrow g - f \geq 0$ so $\int g - \int f \geq 0$.

4) Follows from 3.

1) Let $f, g \in \mathcal{M}(\mathbb{R})$. Write $f = f^+ - f^-$
 $g = g^+ - g^-$ (all are positive)
 $f + g = h^+ - h^-$

$$h^+ - h^- = f + g = f^+ - f^- + g^+ - g^-$$

$$h^+ + f^- + g^- = f^+ + g^+ + h^-$$

By linearity for positive functions

$$\int h^+ dx + \int f^- dx + \int g^- dx = \int f^+ dx + \int g^+ dx + \int h^- dx$$

$$\begin{aligned} \int f + g &= \int h^+ - \int h^- = \int f^+ - \int f^- + \int g^+ - \int g^- \\ &= \int f + \int g. \end{aligned}$$

Scalar mult again clean.

8.iii) Some Inequalities : Assume $f \geq 0$.

[8.4]

Lemma 8.9 : Let $L_{f,t} = \{x \in E \mid f(x) \geq t\}$, $t \geq 0$.

$$\chi_{L_{f,t}} = \begin{cases} 1 & x \in L_{f,t} \\ 0 & \text{else.} \end{cases}$$

Then

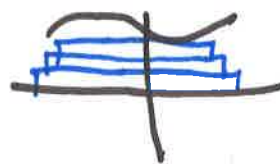
$$f(x) = \int_0^\infty \chi_{L_{f,t}}(x) dt.$$

Proof : Observe

$$\chi_{L_{f,t}}(x) = \begin{cases} 1 & \text{if } f(x) \geq t \\ 0 & \text{if } f(x) < t \end{cases} = \chi_{[0, f(x)]}(t).$$

$$\int_0^\infty \chi_{L_{f,t}}(x) dt = \int_0^\infty \chi_{[0, f(x)]}(t) dx = \int_0^{f(x)} 1 dt = f(x).$$

Corollary 8.10 (Layer Cake Representation)



$$\int_0^\infty f dx = \int_0^\infty m(\{x \mid f(x) \geq t\}) dt$$

Proof :

$$\begin{aligned} \int_0^\infty f(x) dx &= \int_0^\infty \int_0^\infty \chi_{[0, f(x)]}(t) dt dx \\ &= \int_0^\infty \int_0^\infty \chi_{L_{f,t}}(x) dt dx \\ &= \int_0^\infty \int_0^\infty \chi_{L_{f,t}}(x) dx dt \\ &= \int_0^\infty m(L_{f,t}) dt \end{aligned}$$

Pretend we can switch order, we will prove this later.

Thm 8.11 : (Markov's Inequality) If $f \geq 0$,

$$m(\{x \mid f \geq a\}) \leq \frac{\int_E f dx}{a}$$

Proof : $a \cdot m(\{x \mid f \geq a\}) \leq \int f(x) dx$ is obvious.

Thm 8.12 (Chebyshev's Inequality) If $\int |f|^p dx < \infty$ for $p > 0$,

$$m(\{x \mid f \geq a\}) \leq \frac{1}{a^p} \int_E |f|^p dx$$

Proof : Markov applied to $|f|^p$

□

Lecture 9 Lebesgue Integration III: the convergence theorems.

9.1

Question 9.1: Suppose $f_n \rightarrow f$ pointwise a.e. When can we conclude

$$\lim_{n \rightarrow \infty} \int_E f_n dx = \int_E \lim_{n \rightarrow \infty} f_n dx ?$$

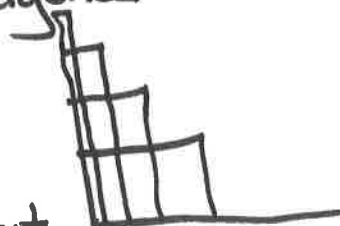
Ex 9.2: Two phenomena can ruin convergence

1) Mass escapes vertically

$$f_n = n \cdot \chi_{[0, 1/n]}$$

Then $f_n \rightarrow 0$ pointwise a.e., but

$$\lim_{n \rightarrow \infty} \int_0^1 f_n dx = \lim_{n \rightarrow \infty} 1 = 1 \neq 0 = \int_0^1 f dx.$$

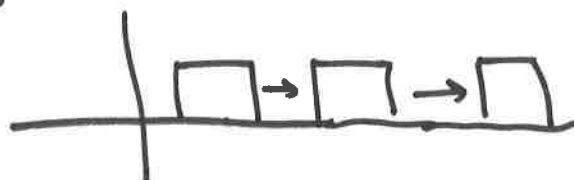


2) Mass escapes horizontally

$$f_n = \chi_{[n, n+1]}$$

Then $f_n \rightarrow 0$ pointwise,

$$\lim_{n \rightarrow \infty} \int_{\mathbb{R}} f_n dx = \lim_{n \rightarrow \infty} 1 \neq 0 = \int_{\mathbb{R}} f dx.$$



Rem 9.3: Recall convergence holds if $m(E) < \infty$ and $|f_n| \leq M$ as it rules out both these behaviors, i.e. Ex 9.2 is "all" that can go wrong.

9.1 Fatou's Lemma

Lemma 9.4 (Fatou's Lemma) Suppose $f_n \geq 0$ are non-negative measurable functions on E . If $f_n \rightarrow f$ pointwise a.e.,

$$\int_E f \leq \liminf_n \int_E f_n.$$

Proof: Note f is measurable since it is pointwise a.e. limit, $f \geq 0$. Let h be bounded, finite support with $0 \leq h \leq f$.

Define

$$h_n = \min \{h, f_n\} \quad \text{on } E \supseteq \text{supp } h.$$

9.2

Then
and

$$0 \leq h_n \leq h \leq M := \text{Sup } h, \text{ so is bounded,}$$
$$\text{supp } (h_n) \subseteq \text{supp } h.$$

By Bounded Convergence Thm (7.9),

$$\lim_{n \rightarrow \infty} \int_E h_n = \lim_{n \rightarrow \infty} \int_{\text{supp } h} h_n = \int_{\text{supp } h} h = \int_E h.$$

But $h_n \leq f_n \quad \forall n$ so

$$\int_E h = \lim_{n \rightarrow \infty} \int_E h_n \leq \liminf_{n \rightarrow \infty} \int_E f_n$$

If $\liminf = \infty$ trivial,
if $\liminf < \infty$ take N st
 $\inf_{n \geq N} = \text{Inf} + \epsilon/2$
increase n so that $\lim \int h_n$ within $\epsilon/2$

Now take $\sup_{h \leq f, \text{supp } h \text{ finite}}$

□

Rem 9.5 : Ex 9.2 shows inequality may be strict w/ right side including the "escaped" mass.

9.ii Monotone Convergence :

Theorem 9.6 (Monotone Convergence) Suppose f_n is an increasing sequence of non-negative functions on E , and $f_n \rightarrow f$ pointwise a.e. Then

$$\lim_{n \rightarrow \infty} \int_E f_n dx = \int_E \lim_{n \rightarrow \infty} f_n dx.$$

Proof : By Fatou,

$$\int_E f \leq \liminf_{n \rightarrow \infty} \int_E f_n dx$$

But

$$\leq \limsup_{n \rightarrow \infty} \int_E f_n dx \leq \limsup_{n \rightarrow \infty} \int_E f dx = \int_E f dx.$$

Therefore $\liminf = \limsup = \int_E f dx.$

□

9.iii) Dominated Convergence

[9.3]

Def 9.7: A measurable function g is said to dominate a sequence f_n if $|f_n| \leq g$ on E for all n .

Lemma 9.8 (Modulus of Integrability) Suppose $f \geq 0$ is integrable. Then $\forall \varepsilon > 0$, $\exists \delta$ such that $m(A) < \delta \Rightarrow \int_A f dx < \varepsilon$.

Proof: Let $\varepsilon > 0$. For m sufficiently large, $\exists f_m$ with $|f_m| \leq M$ and

$$\int_E |f - f_m| dx < \frac{\varepsilon}{2}.$$

Now let $A \subseteq E$ be measurable w/ $m(A) < \delta := \frac{\varepsilon}{2M}$.

Then

$$\begin{aligned} \left| \int_A f dx \right| &\leq \int_A |f| dx \leq \int_A |f - f_m + f_m| dx \\ &\leq \int_A |f - f_m| dx + \int_A |f_m| dx \\ &\leq \int_E |f - f_m| dx + m(A) \cdot M \\ &\leq \frac{\varepsilon}{2} + \frac{\varepsilon}{2} \quad \square \end{aligned}$$

Corollary 9.9: $f: \mathbb{R} \rightarrow \mathbb{R}$ integrable,
 $F(x) = \int_{-\infty}^x f(t) dt$
is uniformly continuous.

Thm 9.10 (Lebesgue Dominated Convergence) Suppose $f_n \rightarrow f$ pointwise a.c. with $|f_n|, |f|$ dominated by g with $\int_E g dx < \infty$. Then $\lim_{n \rightarrow \infty} \int_E f_n dx = \int_E \lim_{n \rightarrow \infty} f_n dx$

Proof: By Lemma 9.8, given $\varepsilon > 0 \exists \delta$ such that $A \subseteq E$ with $m(A) < \delta \Rightarrow \int_A g dx < \varepsilon/2$.

Also $\exists h \leq g$ supported on $[-m, m]$ st $\int g - h dx < \frac{\varepsilon}{2}$.

By Littlewood 3, $\exists A$ with $m(A) < \delta$ s.t. $f_n \rightarrow f$ uniformly outside A .

Then

$$\begin{aligned}
 \left| \int_E f_n - f \, dx \right| &\leq \int_E |f_n - f| \, dx \\
 &\leq \int_{E \setminus [-m, m] \cap E} |f_n - f| \, dx + \int_{[-m, m] \cap E} |f_n - f| \, dx + \int_A |f_n - f| \, dx
 \end{aligned}$$

Note m is defined by g so ind. of n !

$$\begin{aligned}
 &\leq 2 \int_{E \setminus [-m, m] \cap E} |g| \, dx + \int_{[-m, m] \cap E} |f_n - f| \, dx + 2 \int_A |g| \, dx \\
 &\leq 2 \cdot \frac{\varepsilon}{2} + 2m \cdot \sup_{\text{uniform}} |f_n - f| + 2 \cdot \frac{\varepsilon}{2} \\
 &\leq \varepsilon + \varepsilon \quad \text{for } n \text{ large} \quad \square.
 \end{aligned}$$

9. iv) Examples

Ex 9.11 : Evaluate $\lim_{k \rightarrow \infty} \int_0^1 \frac{k[\sin(kx) + \sin(\frac{1}{k}x)]}{\tan(\frac{\pi}{2k}x) + K^{3/2}\sqrt{x}} \, dx$.

$$\left| \frac{k[\sin(kx) + \sin(\frac{1}{k}x)]}{\tan(\frac{\pi}{2k}x) + K^{3/2}\sqrt{x}} \right| \leq \frac{2k}{\text{positive} + K^{3/2}\sqrt{x}} \leq \frac{1}{\sqrt{kx}} \leq \frac{1}{\sqrt{x}}$$

and $\int_0^1 \frac{1}{\sqrt{x}} \, dx = 2x^{1/2} \Big|_0^1 = 2 < \infty$. So

$$\lim_{k \rightarrow \infty} \int_0^1 \frac{k[\sin(kx) + \sin(\frac{1}{k}x)]}{\tan(\frac{\pi}{2k}x) + K^{3/2}\sqrt{x}} \, dx = \int_0^1 \lim_{k \rightarrow \infty} \frac{k[\sin(kx) + \sin(\frac{1}{k}x)]}{\tan(\frac{\pi}{2k}x) + K^{3/2}\sqrt{x}} \, dx = 0.$$

Ex 9.12 : Suppose $f(t, x)$ is C^1 a.e. and $|f(t, x)| \leq m$. Then

$$\frac{d}{dt} \int_{\mathbb{R}} f(t, x) \, dx = \int_{\mathbb{R}} f'(x, t) \, dx$$

Proof : $|f|$ dominates the difference quotients $\frac{f(t+h) - f(t)}{h}$.

Corollary 9.13 : If $E_1 \subseteq E_2 \subseteq \dots$ and $\int_E |f| < \infty$, then

$$\int_{\bigcup E_n} f = \lim_{n \rightarrow \infty} \int_{E_n} f$$

and

$$\int_{\bigcap E_n} f = \lim_{n \rightarrow \infty} \int_{E_n} f. \quad (\text{Apply LDC to } \chi_{E_n} f.)$$

Lecture 10 Differentiation and Integration I: monotone functions 10.1

Question 10.1: Under what assumptions of f does the FTC

$$\frac{d}{dx} \int_a^x f(t) dt = f(x)$$

hold? Note we saw that it fails for $\sin(x)$ the Cantor function.

Ex 10.2: There exists a continuous but nowhere differentiable function. Let $a_n = 10^{-n}$, $b_n = 10^{2n}$.

Consider

$$f = \sum_{n=1}^{\infty} a_n \sin(b_n x) \quad \text{on } [0, 2\pi].$$

f is a uniform limit of continuous, so continuous,

but

$$f'(x) = \sum_{n=1}^{\infty} a_n b_n \sin(b_n x) = \sum_{n=1}^{\infty} 10^n \sin(b_n x).$$

would have to hold. Can show $\frac{f(x+h) - f(x)}{h} \rightarrow \infty$ everywhere.

10.2) Monotone Functions

Def 10.3: A function $f: [a, b] \rightarrow \mathbb{R}$ is said to be monotone if $x \leq y \Rightarrow f(x) \leq f(y)$.

Prop 10.4: Let $f: [a, b] \rightarrow \mathbb{R}$ be monotone. Then f is continuous away from a countable (hence measure 0) set.

Proof: Suffices to assume $[a, b]$ finite and take unions.

$$\text{Let } f^-(x_0) = \sup \{ f(x) \mid a < x < x_0 \}$$

$$f^+(x_0) = \inf \{ f(x) \mid x_0 < x < b \}.$$

Discontinuity iff $f^- < f^+$ is strict. If so $[f^-(x_0), f^+(x_0)] \subseteq [f(a), f(b)]$.

$\forall n \in \mathbb{N}$, only finitely many have size $> \frac{1}{n}$. So

$$\text{discontinuities} = \bigcup_{n \in \mathbb{N}} \{\text{finite}\}.$$

□

Ex 10.5 : If $C \subseteq [a, b]$ is countable $\exists f: [a, b] \rightarrow \mathbb{R}$ monotonic 10.2
with discontinuities at C .

Enumerate $C = \{q_1, q_2, \dots\}$

$$f(x) = \sum_{i: q_i \leq x} \frac{1}{2^i}.$$

Thm 10.6 (Monotone Differentiability) If $f: [a, b] \rightarrow \mathbb{R}$ is monotone it is differentiable almost everywhere.

10.ii) Covering Lemmas

Lemma 10.7 : Let $K \subseteq \mathbb{R}^n$ be compact. If $\mathcal{U} = \bigcup_{i \in I} D_i$ is a cover by balls, $\exists i \in 1, \dots, N$ st.
$$K \subseteq \bigcup_{i=1}^N 3D_i$$

Proof : Let D_i for $1 \dots N$ be a finite subcover ordered by decreasing radii. $D_i = B(x_i, r_i)$ $r_1 \geq r_2 \geq \dots$.
Starting from $i=1$ throw out each D_i if it is not disjoint from those already selected. Suppose $x \in K$. Either x is in chosen D_i or another D_j . In the latter case $\exists i$ so $x \in D_i \cap D_j$ w/ $i \geq j$, but then
 $x \in 3D_i$.  □

Def 10.8 : A collection $\mathcal{U} = \{D_i\}$ of balls is a Vitali covering of K if $\forall x, \forall \varepsilon > 0$, $\exists i \in I$ such that $x \in D_i$ and $r_i < \varepsilon$.

Lemma 10.9 (Vitali Covering Lemma) Let $E \subseteq \mathbb{R}^n$ be a set of finite positive measure, and let $\mathcal{U}$ be a Vitali covering. Then $\exists$ finite collection $D_i \in \mathcal{U}$ st.
$$m(E \Delta \bigcup_{i=1}^N D_i) < \varepsilon.$$

Lemma 10.10 : Let $\mathcal{U}$ be a Vitali covering of K compact. Then 10.3

$$\forall N \in \mathbb{N} \exists D_i \text{ disjoint so}$$

$$K \subseteq \bigcup_{i=1}^N \bar{D}_i \cup \bigcup_{i=N+1}^{\infty} 3D_i$$

Proof : Assume $\mathcal{U} = \bigcup_{n \in \mathbb{N}} B(x_n, \frac{1}{n})$ for x_n finite, by compactness.

Proceed w/ algorithm as before on (now countable sequence).

Either $x \in \bigcup_{i=1}^N \bar{D}_i$ or it lies in a ball thrown out, and radii $r_i \geq r_j$, so same logic. $\square$

Proof (of Vitali Covering Lemma) We may choose

$$K \subseteq E \subseteq \mathcal{U} \quad \text{and } m(\mathcal{U} \setminus K) < \varepsilon.$$

$\nearrow$ closed, finite measure $\Rightarrow$ compact $\nwarrow$ open

By 10.10, $K \subseteq \bigcup_{i=1}^N \bar{D}_i \cup \bigcup_{i=N+1}^{\infty} 3D_i \quad \forall N,$

By throwing out balls of large radii, may assume each $D \in \mathcal{U}$ lies in $\mathcal{U}$.

Since D_i disjoint, $\sum_{i=1}^{\infty} m(D_i) < m(\mathcal{U}) < m(E) + \varepsilon.$

Let N be large enough so that $\sum_{i=N+1}^{\infty} 3m(D_i) < \varepsilon.$

$$m(E \setminus \bigcup_{i=1}^N D_i) \leq m(\mathcal{U} \setminus \bigcup_{i=1}^N D_i) \leq m(\mathcal{U} \setminus K) + m(K \setminus \bigcup_{i=1}^N \bar{D}_i) \leq m(\mathcal{U} \setminus K) + m(K \setminus \bigcup_{i=1}^N \bar{D}_i)$$

$$m(\bigcup_{i=1}^N D_i \setminus E) \leq \varepsilon + m(\bigcup_{i=1}^N D_i \setminus K) \leq 2\varepsilon \text{ since } \mathcal{U} \text{ covers } K. \quad \square$$

Proof (of monotone differentiability)

Suffices to assume $[a, b]$ finite, and f continuous (bc discords are measure 0)

For $x \in [a, b]$, let $I_{\delta_1, \delta_2} = [x - \delta_1, x + \delta_2].$

Set $D_{\delta_1, \delta_2}(x) = \frac{f(x + \delta_2) - f(x - \delta_1)}{x + \delta_2 - (x - \delta_1)}.$

Clearly $\lim_{\substack{\delta_1 \rightarrow 0 \\ \delta_2 \rightarrow 0}} D_{\delta_1, \delta_2}(x)$ exists $\Rightarrow f'(x)$ exists by monotonicity by continuity.

Thus if f' does not exist $\lim_{\delta, \delta_2 \rightarrow 0} D_\delta(x)$ does not exist, (10.4)
and $\exists$ intervals I, J so that

$$D_I(x) \neq D_J(x) \Rightarrow \exists \text{ a rational in between.}$$

Let $E_{r,s}$ be $\{x \mid \nexists r, s \in \mathbb{Q} \text{ w/ } D_I(x) < r < s < D_J(x)\}$
 $\exists I, J$ arbitrarily small.

Thus if $f'(x)$ DNE, $x \in E_{r,s}$ for some r, s .

Claim: $m(E_{r,s}) = 0$ (hence $m(\cup E_{r,s}) = m\{x \mid f' \text{ DNE}\} = 0$).

Idea: Intuitively, $f' > s$, and $f' < r$ both hold, so we show
 $s \cdot m(A) \leq m(f(E)) \leq r m(A) \rightarrow$ since $r \neq s$
if $m(A) \neq 0$.

Let $\varepsilon > 0$. By Vitali's Lemma $\exists \bigcup_{n=1}^N I_n$ disjoint st
 $m(E_{r,s} \Delta \bigcup_{n=1}^N I_n) < \varepsilon$.

In fact, can assume $D_{I_n}(x) < r$. (Take $\mathcal{U}$ to be the covering of $E_{r,s}$ by the intervals I of size $\frac{1}{n}$).

$$\text{Thus } \sum |f(I_n)| \leq r \sum |I_n|$$

Now apply Vitali to $E_{r,s} \cap \bigcup_{n=1}^m I_n$ using covering by $\mathcal{J}_s$.

$$\exists \bigcup_{m=1}^M J_m \text{ st } m(E_{r,s} \cap \bigcup_{n=1}^N I_n \Delta \bigcup_{m=1}^M J_m) < \varepsilon \text{ and } D_{J_m}(x) > s.$$

$$\Rightarrow \sum |f(J_m)| \geq s \sum |J_m|$$

$\rightarrow$ monotonicity $\bigcup J_m \subseteq \bigcup I_n$ and disjoint

$$s m(\bigcup J_m) \leq \sum |J_m| \leq \sum |f(J_m)| \leq \sum |f(I_n)| \leq r \sum |I_n| = r m(\bigcup I_n)$$

$$|m(E_{r,s}) - m(\bigcup I_n)| \leq \varepsilon \Rightarrow \lim \varepsilon \rightarrow 0$$

$$|m(E_{r,s}) - m(\bigcup J_m)| \leq 2\varepsilon \Rightarrow s m(E_{r,s}) \leq r m(E_{r,s}) \rightarrow \text{unless } m(E_{r,s}) = 0$$

Lecture 11 Integration and Differentiation II: absolute continuity 11.1

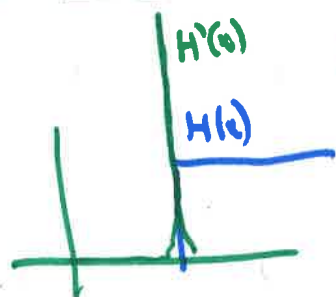
Recall if $F: [a, b] \rightarrow \mathbb{R}$ is measurable and monotone, then $F' = f$ exists a.e.

Ex 11.1: As before if $C(x): [0, 1] \rightarrow \mathbb{R}$ is the Cantor function, $C' = f$ indeed exists and is 0 a.e.
Hence

$$0 = \int_0^1 C'(x) \neq C(1) - C(0) = 1.$$

But $\leq$ is true.

Intuition 11.2: Consider $H(x) = \begin{cases} 0 & x \leq 1/2 \\ 1 & x > 1/2 \end{cases}$

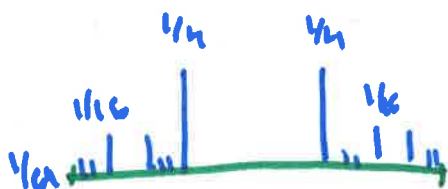


Then intuitively $H'(x) = \delta(x - 1/2)$ is the derivative

$$\int_a^b H'(x) = \begin{cases} 1 & \text{if } a \leq 1/2 \leq b \\ 0 & \text{else.} \end{cases}$$

For the Cantor function, increase comes

from $\sum_{C \in C} \frac{1}{4^k} \delta_{C_k}$ $k = \text{layer of Cantor set.}$



Thm 11.3: If $f: [a, b] \rightarrow \mathbb{R}$ is monotone increasing

$$\int_a^b f'(x) dx \leq f(b) - f(a).$$

Proof: Let $f_n(x) = \frac{f(x + 1/n) - f(x)}{1/n}$. By f' existing a.e. f extend f by $f(a), f(b)$ const.

$f_n(x) \rightarrow f'(x)$ pointwise a.e.

Thus

$$\int f' \leq \liminf \int f_n \leq \limsup \int f_n$$

$$\int_a^b f_n = \frac{1}{n} \int_{a+1/n}^{b+1/n} f(x) dx - \frac{1}{n} \int_a^b f(x) dx = \frac{1}{n} \int_b^{b+1/n} f(x) dx - \frac{1}{n} \int_{a+1/n}^{a+1/n} f(x) dx \leq f(b) - f(a)$$

by monotone.

11.ii) Absolute Continuity + Bounded Variation

11.2

Q 11.4 : What is the condition for equality? We have to rule out the δ -function jumps. We will define

$$\begin{aligned} BV[a,b] &\supseteq AC[a,b] \\ &= \{f \text{ has bounded variation}\} = \{f \text{ absolutely continuous}\} \end{aligned}$$

Def 11.5 : A function has bounded variation if

$$\|f\|_{BV} := \sup_{P \in [a,b]} \sum_{i=0}^n |f(a_{i+1}) - f(a_i)|$$

is finite, where the sup is over all finite partitions

$$a = a_0 \leq a_1 < a_2 < \dots < a_n = b.$$

Prop 11.6 : A function f has bounded variation iff

$$f = g(x) - h(x)$$

for g, h monotone increasing.

Proof : Suppose $f = g(x) - h(x)$. Then

$$\begin{aligned} \|f\|_{BV} &= \sum_{k=0}^n |g(x_{k+1}) - g(x_k) + h(x_{k+1}) - h(x_k)| \\ &\leq \sum_{k=0}^n |g(x_{k+1}) - g(x_k)| + |h(x_{k+1}) - h(x_k)| \\ &\leq g(b) - g(a) + h(b) - h(a) < \infty \end{aligned}$$

Now suppose $\|f\|_{BV} < \infty$. Write $y = y^+ - y^-$ where $y^\pm \geq 0$.
 $y = y^+ - y^-$ for $y^\pm \geq 0$.

$$f_{\pm}(x) = \sup_P \sum_{k=0}^n [f(a_{k+1}) - f(a_k)]_{\pm}$$

where P partitions $[a, x]$.

Then $f_{\pm}$ are monotone and bounded since $\|f\|_{BV} < \infty$. 11.3

Claim $f = f_+ - f_- + f(a)$.

Note $f_{\pm}$ are increasing as we refine partition (by triangle ineq)
so can find P subdivision for which both are within $\varepsilon/2$ of sup.

Then

$$\sum_{i=1}^n [f(a_i) - f(a_{i-1})]_+ - [f(a_{i-1}) - f(a_i)]_-$$
$$= \sum_{i=1}^n f(a_{i-1}) - f(a_i) = f(a_n) - f(a_0) = f(x) - f(a) + \varepsilon.$$

Corollary 11.7: If $f \in BV[a, b]$ then f' exists a.e.
and $f' \in L^1[a, b]$

Proof: apply monotone differentiability to g, h . □

Def 11.8: A function $F: [a, b] \rightarrow \mathbb{R}$ is absolutely continuous
if $\forall \varepsilon > 0, \exists \delta > 0$ st. for any $\sqcup I_n$ w/

$$\sum |I_n| < \delta, \quad \sum_{i=1}^n |f(a_i) - f(b_i)| < \varepsilon.$$

For $I_j = (a_j, b_j)$.

Prop 11.9: Suppose $f: [a, b] \rightarrow \mathbb{R}$ is integrable. Then
 $F(x) = \int_a^x f \, dt$ is absolutely continuous.

Proof: $\forall \varepsilon > 0 \exists \delta$ s.t. if $\mu(A) < \delta$ then $\int_A |f| \, dx < \varepsilon$.

(approximate by bounded)

Ex 11.10 : Absolute continuity is designed to precisely rule out the Cantor function. If f has discontinuity jumps

11.4



then $\bigcup I_n$ can be taken $< \varepsilon$ while $|f(b) - f(a)| > \text{const}$

Eg. 2^n intervals of length 3^{-n} so that

$$\sum |f(I_n)| = 1 \text{ but } |I_n| = \left(\frac{2}{3}\right)^n \rightarrow 0.$$

Prop 11.11 : Recall Lipschitz means $\exists M$ st

$$|f(b) - f(a)| \leq M|b - a|,$$

One has $\swarrow$ Lipschitz

$$C^0[a, b] \subset AC[a, b] \subset BV[a, b]$$

Proof : Let $\varepsilon > 0$, and δ be given by A.C. Then

$$\|f\|_{BV} \leq \varepsilon \frac{(b-a)}{\delta}$$

For Lipschitz, take $\delta = \frac{\varepsilon}{M}$

Corollary 11.12 : If $f \in AC[a, b]$ then f' exists a.e. D

11.iii) The FTOC

Theorem 11.13 (FTOC with absolute continuity)

1) Suppose $f \in L^1[a, b]$. Then $F = \int_a^x f \, dt$ is A.C. and

$$\frac{d}{dx} F = f \quad (\text{a.e. hence in } L^1).$$

2) Suppose $f \in AC[a, b]$. Then

$$\int_a^x f'(t) \, dt = f(x) - f(a).$$

Proof (next time)

Lecture 12 Integration and Differentiation III: generalized FTC.

Recall we wish to show

Thm 11.13 (generalized FTC)

1) If $f \in L^1[a, b]$, then $F(x) = \int_a^x f \, dt \in AC[a, b]$

and $F'(x) = f$ a.e.

2) If $f \in AC[a, b]$, then f' exists a.e. and
 $\int_a^x f'(t) \, dt = f(x) - f(a)$.

Rem 12.1: Phrased a different way,

$$\begin{array}{c} D = \frac{d}{dx} \\ AC[a, b] / \mathbb{R} \xrightarrow{\sim} L^1[a, b] \\ \uparrow \\ I = \int_a^x dt \end{array}$$

is an isomorphism

12.1) The lemmas

Intuition 12.1: We will show $D \circ I$ and $I \circ D$ give the identity.

$\Rightarrow D \circ I$ first prove for Lipschitz, then take limits
 $I \circ D$ shown to be injective.

Lemma 12.3: If $I(f) = 0$ for $f \in L^1[a, b]$, then $f = 0$, i.e.
 I is injective.

Proof: If $f \neq 0$ in L^1 , then $\exists E \subseteq [a, b]$ s.t. $f \neq 0$ on E
 $\wedge m(E) > 0$. Take $F \subseteq E$ closed of positive measure.
 F is a union of intervals, take one, J . $\int_J f \neq 0$ so
 $J = [c, d]$ cannot have $\int_c^d f \, dt = \int_a^d f \, dt = 0$ $\square$

Lemma 12.4 (Lipschitz Case) Suppose $f \in L^\infty[a, b]$. Then $I(f)$ is Lipschitz, and $D \circ I(f) = f$ a.e. 12.2

Rem 12.5: This shows the theorem on the subspaces

$$C^0[a, b] \simeq L^\infty[a, b]$$

$$\cap AC[a, b] \simeq \cap L^1[a, b]$$

Proof: Let $|f| \leq M = \|f\|_{L^\infty}$. Then $F = I(f)$ satisfies

$$|F(x+h) - F(x)| = \left| \int_x^{x+h} f(t) dt \right| \leq M \cdot h$$

is Lipschitz. Let $F_n = \frac{1}{n} [F(x + 1/n) - F(x)] \leq M$

and $F_n \rightarrow F'$ pointwise a.e. so by dominated convergence.

$$\begin{aligned} \int_a^x F'(t) dt &= \lim_{n \rightarrow \infty} \int_a^x n [F(t + 1/n) - F(t)] dt \\ &= \lim_{n \rightarrow \infty} \int_{a+1/n}^{x+1/n} n F(t) dt - \int_a^{x+1/n} n F(t) dt \\ &= \lim_{n \rightarrow \infty} \int_x^{x+1/n} n F(t) dt - \int_a^{a+1/n} n F(t) dt = F(x) - F(a) \end{aligned}$$

Thus $\int_a^x F'(t) - f(t) dt = 0 \quad \forall x$, so $F' = f$ a.e. by 12.3. $= \int_a^x f(t) dt$

Lemma 12.6: Let X, Y be sets and $X \xrightarrow{D} Y$ maps such that $\xrightarrow{D} \xrightarrow{I} = \text{id}_Y$. If D is injective, then $I \circ D = \text{id}_X$.

Proof: $D(I(D)x) = Dx$, hence $I \circ D x = x$ by injectivity.

12.ii) The proof

12.3

Proof $I \circ D = \text{id}$ (general case) Suppose $f \in L^1[a, b]$

Write $f = f^+ - f^-$, so enough to assume $f \geq 0$.

Let $f_n = \min(n, f)$, hence $f_n \rightarrow f$ pointwise and is monotone.

Also $F_n = I(f_n)$ is monotone and $F_n \rightarrow F$ pointwise.

By previous lemma,

$$F_n'(x) = f_n \quad \text{a.e.}$$

Therefore

$$\begin{aligned} \int_a^x F' dt &\geq \int_a^x f_n dt = F_n(x) - F_n(a) \\ \int_a^x F' dt &\geq \lim_{n \rightarrow \infty} F_n(x) - F_n(a) \\ &= F(x) - F(a) = \int_a^x f dt = I(f). \end{aligned}$$

$$\int_a^x F' dt \leq F(x) - F(a) = I(f) \quad \text{b.c. } F \text{ is monotone.}$$

Hence $I(F') = I(f)$, and $F' = f$ a.e. by injectivity.

Proof that D is injective). Suppose $F \in AC[a, b]$ and $F'(x) = 0$ a.e. then F is constant.

Let $\varepsilon > 0$, δ as in def of a.e., may assume $\delta < \varepsilon$.

Let $B = \left\{ U \subseteq [a, b] \text{ open st } \left| \frac{F(u)}{|u|} \right| < \frac{\delta}{n} \right\}$ for all n .

Since $F'(x) = 0$, and this is open, it is a covering of the set where F' exists.

Then let $I_1, \dots, I_n$ be intervals st. $m([a, b] - I) < \delta$.

$$\text{Then } |F(x) - F(a)| \leq \varepsilon + \sum_{I_n} |F(I_n)|$$

$$\leq \varepsilon + \sum \delta |I_n| \leq \varepsilon (1 + (b-a)) \rightarrow 0. \quad \square$$

12.4 ii) : Rectifiable Curves

Let $\gamma = (x(t), y(t))$ be such that

$$\|\gamma\|_{BV} < \infty.$$

Def 12.7 : A curve satisfying the above is said to be rectifiable.

Its length is

$$L(\gamma) = \int_a^b |x'(t)^2 + y'(t)^2|^{1/2} dt < \infty.$$

Def 12.8 : Let $K_\delta = \{(x, y) \mid \inf_{t \in [a, b]} d((x, y), \gamma(t)) < \delta\}$.

The 1-dim Minkowski content of γ is $\lim_{\delta \rightarrow 0} \frac{m(K_\delta)}{2\delta}.$

Ex 12.9 : The Koch snowflake is not rectifiable.

• A space filling curve has ∞ Minkowski content.

Rem 12.10 : Minkowski content leads to notion of "Hausdorff measure".

This leads to "Geometric measure theory" and is key for studying singular limiting behavior of PDEs/minimal surfaces.

Thm 12.11 (Isoperimetric Inequality)

Let $\Omega \subset \mathbb{R}^2$ be measurable, with $\partial\Omega$ rectifiable.

Then $4\pi m(\Omega) \leq L(\partial\Omega)^2,$

with equality iff it's a disk.

Let X be a metric space.

Def 13.1 : An extension measure on X is a function $\mu^*: 2^X \rightarrow [0, \infty]$ with

- 1) $\mu^*(\emptyset) = 0$
- 2) $E_1 \subseteq E_2 \Rightarrow \mu^*(E_1) \leq \mu^*(E_2)$
- 3) $\mu^*\left(\bigcup_{j=1}^{\infty} E_j\right) \leq \sum_{j=1}^{\infty} \mu^*(E_j)$.

* A set $E \subseteq X$ is said to be measurable if $\mu^*(A) = \mu^*(A \cap E) + \mu^*(A \cap E^c)$ for all $A \subseteq X$.

Prop 13.2 : $\mathcal{M}(X) = \{E \subseteq X \mid E \text{ is measurable}\}$ forms a σ -algebra, and $\mu^*|_{\mathcal{M}(X)} : \mathcal{M}(X) \rightarrow \bar{\mathbb{R}}^{\geq 0}$ is a measure.

Proof : Nothing about previous proof used $\mathbb{R}$!

Def 13.3 : A pre-measure on an algebra $\mathcal{A} \subseteq 2^X$ (closed under finite $\cup, \cap$) is a function $\mu_0 : \mathcal{A} \rightarrow [0, \infty]$ such that

- 1) $\mu_0(\emptyset) = 0$
- 2) $\mu_0\left(\bigsqcup_{k=1}^{\infty} E_k\right) = \sum_{k=1}^{\infty} \mu_0(E_k)$

the outer measure completing a premeasure is

$$\mu^*(E) = \inf \left\{ \sum_{j=1}^{\infty} \mu_0(E_j) \mid E \subseteq \bigcup E_j, E_j \in \mathcal{A} \right\}$$

Prop 13.4 : μ^* as above is an extension measure, and $\mathcal{A} \subseteq \mathcal{M}(X)$ in the induced measure.

Thm 13.5 : If $\mathcal{A} \subseteq 2^X$ is a (covering) algebra ^{has countable cover of μ_0 sets.} and μ_0 a pre-measure, then $\exists!$ measure μ extending μ_0 .

Proof : Existence immediate from 13.2 + 13.4. For uniqueness Suppose ν, μ are two measures extending μ_0 .

$$\text{If } F \subset \bigcup E_j \text{ w/ } E_j \in \mathcal{A},$$

$$\nu(F) \leq \sum_{j=1}^{\infty} \nu(E_j) = \sum_{j=1}^{\infty} \mu_0(E_j)$$

$$\nu(F) \leq \inf (\quad) = \mu(F).$$

and vice-versa. □

Proof of 13.4 : μ is clearly an outer measure

b/c

$$1) \mu(\emptyset) = 0, \quad 2) E_1 \subset E_2 \text{ means } \mu_0(E_1 \cup E_2 \setminus E_1) = \mu(E_1) + \underbrace{\quad}_{\geq 0}$$

in $\mathcal{A}$.
so,

$$\inf \{ \text{covers of } E_1 \} \leq \inf \{ \text{covers of } E_2 \}$$

b/c strictly bigger set.

$$3) \mu^*\left(\bigcup_{j=1}^{\infty} E_j\right) = \mu^*\left(\bigcup_{j=1}^{\infty} (E_j \setminus \bigcup_{k=1}^{j-1} E_k)\right) \leq \sum_{j=1}^{\infty} \mu(E_j \setminus \bigcup_{k=1}^{j-1} E_k) \leq \sum_{j=1}^{\infty} \mu(E_j).$$

On $\mathcal{A}$, $\mu(E) \leq \mu_0(E)$ since it covers itself

and if $E \subset \bigcup_{j=1}^{\infty} E_j$ then

$$\mu_0(E) = \sum (E_j \setminus \bigcup_{k=1}^{j-1} E_k) \leq \sum \mu_0(E_k) \quad \text{and take inf.}$$

$\mathcal{A} \subset \mathcal{M}(\mathbb{R})$ is similar. □

13.ii) Lebesgue Measure on $\mathbb{R}^n$

Def 13.6 : Given (X, Σ, μ) , $(Y, \mathcal{M}, \nu)$

the product pre-measure is the measure

$$\lambda(A \times B) = \mu(A) \nu(B)$$

on the algebra of products of measurable sets.

Claim 13.7 : this is a pre-measure.

• $\lambda(\emptyset) = 0 \quad \checkmark$

• If $A \times B = \bigsqcup A_j \times B_j$ all products of measurable,
then (x, y) belongs to a single ~~$A_j \times B_j$~~ $A \times B_j$ for B_j depends
on y .
hence $B = \bigsqcup B_j$.

Then $\chi_A(x_1)\mu_2(B) = \sum_{j=1}^{\infty} \chi_{A_j}(x_1)\mu_2(B_j)$ for $x_1 \in A_j$.

13.3

Integrating

$$\begin{aligned} \mu(A)\mu(B) &= \lim \int \chi_A(x_1)\mu_2(B) dx_1 = \lim_{N \rightarrow \infty} \int \sum_{j=1}^N \chi_{A_j}(x_1)\mu_2(B_j) dx_1 \\ &= \lim_{N \rightarrow \infty} \sum_{j=1}^N \mu(A_j)\mu(B_j). \quad \square \end{aligned}$$

Def 13.8: The product measure on $X \times Y$ is the one induced by the product pre-measure.

Thm 13.9: The product of Lebesgue measures on $\mathbb{R}^k \times \mathbb{R}^m$ is the Lebesgue measure on $\mathbb{R}^{m+k}$.

Proof: $k=m=1$. Suffices to show that $\lambda = m^*$ product = Lebesgue outer

$A \subset \mathbb{Q}; B \in \mathcal{L}(P_k)$ on $A \times B$ measurable.

$\Rightarrow m(A \times B) \leq \sum_{Q_j, P_k} m(Q_j \times P_k) = \sum_{Q_j, P_k} m(Q_j)m(P_k)$ they agree on cubes!

$\leq \sum m(Q)m(P) \leq m(A)m(B) + \varepsilon$.

$\Leftarrow$ $\{Q_k\}$ ^{almost disjoint cubes} cover $A \times B$, then $Q_k = P_j \times M_i$. Therefore

if $(x, y) \in A \times B$, $x \in P_j$, $y \in M_i$ same i so

$A \subseteq \bigcup P_j$ $B \subseteq \bigcup M_i$

$$m(A)m(B) \leq \sum_{j,i} m(P_j)m(M_i) = \sum m(Q_k) \leq m(A \times B) + \varepsilon. \quad \square$$

13.iii) Fubini's Theorem

Thm 13.10 (Fubini I) Suppose that $\int |f| dV < \infty$ on $\mathbb{R}^m \times \mathbb{R}^n$ where dV is the product ^{Lebesgue} measure. Then for a.e. $x \in \mathbb{R}^m, y \in \mathbb{R}^n$

- $f(x, -)$ is integrable on $\mathbb{R}^n$, $f(-, y)$ on $\mathbb{R}^m$; and

$$\int_{\mathbb{R}^{m+n}} f dV = \int_{\mathbb{R}^m} \left(\int_{\mathbb{R}^n} f(x, y) dy \right) dx = \int_{\mathbb{R}^n} \left(\int_{\mathbb{R}^m} f(x, y) dx \right) dy$$

Rem 13.11: Requires verifying $\int |f| dV < \infty$.

13.4

Thm 13.11 (Fubini II / Tonelli)

Suppose $|f| \geq 0$ on $\mathbb{R}^{m+n}$ is measurable. Then $f(x, -)$, $f(-, y)$ are ~~integrable~~ measurable for a.e. x, y resp and

$$\int_{\mathbb{R}^{m+n}} |f| dV = \int_{\mathbb{R}^n} \left(\int_{\mathbb{R}^m} |f| dy \right) dx = \int_{\mathbb{R}^n} \left(\int_{\mathbb{R}^m} f(x, y) dx \right) dy$$

In particular, the hypotheses of Fubini's thm hold.

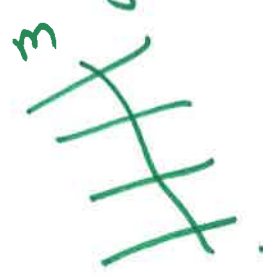
Proof: Stein-Shakarchi pg 76.

Corollary 13.12: If $E \subseteq \mathbb{R}^{m+n}$, then $E_x \cap \{x \times \mathbb{R}^n\}$ is measurable for almost every x , and

$$m(E) = \int_{\mathbb{R}^m} m(E_x) dx = \int_{\mathbb{R}^n} m(E_y) dy.$$

Proof: Tonelli on χ_E .

Corollary 13.13 (informal) If $M \subseteq \mathbb{R}^n$ is a submanifold, w/ $\text{codim } M \geq 1$, then M has measure 0.

Proof: After a change of variable, $\text{Vol}(M) = \int_{\substack{\mathbb{R}^k \times \{0\} \\ \subseteq \mathbb{R}^n}} |\det \psi| dx dy$

$$= \int_{\mathbb{R}^k} \chi_M |\det \psi| dx dy$$

↓

$$= 0 \text{ for a.e. } y \in \mathbb{R}^{n-k}.$$

0

Lecture 14 Radon-Nikodym Theorem

Let $(X, \mathcal{M}, \mu), (X, \mathcal{M}, \nu)$ be two different measures on the same space.

Def 14.1: ν is said to be absolutely continuous with respect to μ if

$$E \in \mathcal{M}(X), \mu(E) = 0 \Rightarrow \nu(E) = 0.$$

Ex 14.2: Let $f \geq 0$ be continuous, bounded.

$$\text{Set } \nu(E) = \int_E f \, d\mu.$$

Clearly $\mu(E) = 0 \Rightarrow \nu(E) = 0$.

Same holds for f measurable (choose bounded approximation).

Lemma 14.3: ν is a.c. with respect to μ iff

$$\forall \varepsilon > 0, \exists \delta \text{ such that}$$

$$\mu(E) < \delta \Rightarrow \nu(E) < \varepsilon.$$

Proof: $\Rightarrow$ a.c. is obvious.

$\Leftarrow$ Suppose not. Then $\exists E_n$ w/ $\mu(E_n) \leq \frac{1}{2^n}$ but $\nu(E_n) \geq \varepsilon_0$.

Let $A_n = \bigcup_{k=n}^{\infty} E_k$, so $A_1 \supseteq A_2 \supseteq \dots$

Then $\mu(A_n) \leq \frac{1}{2^{n-1}}$ $\nu(A_n) \geq \varepsilon_0$

Set $A = \bigcap_{n=1}^{\infty} A_n$. Then $\mu(A) \leq \mu(A_n) \forall n$
 $= 0.$

But by cont. of measure $\nu(A) = \lim \nu(A_n) \geq \varepsilon_0 \rightarrow \Leftarrow$.

Def 14.4 : Two measures are said to be mutually singular if

[14.2]

$\exists A, B$ disjoint s.t.

$$\nu(E) = \nu(A \cap E) \quad \mu(E) = \mu(B \cap E)$$

for all $E \in \mathcal{M}$

Fact 14.5. $L^2(\mathbb{R}, \mu)$ is a complete space in any measure.

Thm 14.6 (Radon-Nikodym)

Suppose μ is (σ -finite) and ν is absolutely continuous wrt μ . Then $\exists f \in L^1_{loc}(\mathbb{R}^n; \mu)$ such that

$$\nu(E) = \int_E f \, d\mu.$$

Rem 14.7 : $f \in L^1_{loc}(X; \mu)$ means $f \in L^1(E; \mu)$ for any set of finite measure.

Ex 14.8 : If $\nu = 2\mu$, then clearly $f = 2$ is not in $L^1(\mathbb{R}^n)$.

Def 14.9 : On $\mathbb{R}$, $\nu(E) = \int_E g \, dx$ with $g \in C^1$, then

~~$\nu(E) = \int_E g \, dx$~~ Set $G(x) = \nu[0, x]$.

Then $G(x) = \int_0^x g \, dx$ so $G' = g$ by FTC.

f is in general called the Radon-Nikodym derivative denoted $\frac{d\nu}{d\mu}$.

Proof: First assume X is a cube! Finite measures for both! [14.3]
 Let $\nu = \mu + \nu$. Let $\varphi \in L^2(\mathbb{R}^n; \nu)^*$ be given by

$$\varphi(f) = \int_{\mathbb{R}^n} f d\nu.$$

Clearly $\|\varphi(f)\|_{\nu} \leq \int_{\mathbb{R}^n} f d\nu \leq \|f\|_{L^2(\nu)}(X)$ so bounded.

By Riesz rep, $\exists g$ st.

$$\int_{\mathbb{R}^n} f d\nu = \int_{\mathbb{R}^n} f g d\mu.$$

Claim: $0 \leq g < 1$ a.e. w.r.t μ .

First, take $f = \chi_E$ for some E so,

$$\int_{\mathbb{R}^n} g d\mu = \int_{\mathbb{R}^n} \chi_E d\nu = \nu(E) < \mu(E) + \nu(E).$$

< because if $\exists E$ w/ $\mu(E) > 0$, then equality implies $\mu(E) = 0$
 $\rightarrow$ absolute cont. $\square$

Now take $f = \chi_E (1 + g + \dots + g^n)$. Since

$$\int_{\mathbb{R}^n} f d\nu = \int_{\mathbb{R}^n} f g (d\nu + d\mu)$$

$$\int_{\mathbb{R}^n} f (1 - g) d\nu = \int_{\mathbb{R}^n} f g d\mu$$

then

$$\int_{\mathbb{R}^n} f (1 - g) d\nu = \int_{\mathbb{R}^n} \chi_E \frac{1 - g^{n+1}}{1 - g} (1 - g) d\nu = \int_{\mathbb{R}^n} \chi_E (1 - g^{n+1}) d\nu = \int_{\mathbb{R}^n} g (1 + g + \dots + g^n) d\mu$$

Now

$1 - g^{n+1} \rightarrow 1$ pointwise, D.C.T. $\nu(E)$

$g + g^2 + \dots \rightarrow \frac{g}{1 - g}$

$\Rightarrow$

$$\int_{\mathbb{R}^n} \chi_E d\nu = \int_{\mathbb{R}^n} f d\mu \quad f = \frac{g}{1 - g}.$$

For general case, cover

$$X = \bigcup E_j$$

with $\mu(E_j) < \infty$, $\nu(E_j) < \infty$ disjoint. Take $f = \sum f_j$. $\square$

Thm 14.10 (Lebesgue Decomposition)

Let $(X, \mathcal{M}, \mu) \subset (\mathbb{R}^n, \mathcal{M}, m)$ be a σ -finite measure space. Let ν on $(X, \mathcal{M}, \nu)$ be another σ -finite measure. Then, $\exists!$ decomposition

$$\nu = \nu_a + \nu_s$$

st. ν_a is absolutely continuous wrt μ , ν_s is mutually sing.

Proof: In above proof, "atomic" part ν_s arises from $\{g=1\}$. Take $B = \{x \in X \mid g(x)=1\}$.

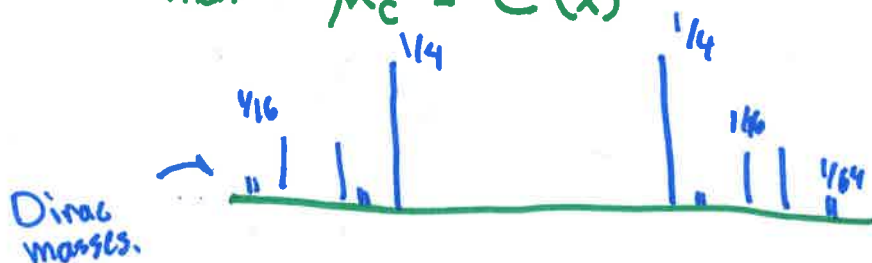
Then $\nu_s(E) = \nu(E \cap B)$.

$\mu(E) \in \nu(E \cap B^c)$ mutually sing.

Ex 14.11: Let μ_c be the measure

$\mu_c([0, x]) = C(x)$ the cantor function

Then $\mu_c = "C'(x)"$



$x \in C$

$x = x_0 x_1 x_2 \dots$

μ_c is mutually sing. w/ Lebesgue b/c $m(C) = 0$.

Lecture 15 General Measures III: Interlude on probability

15.1

Def 15.1: A probability space $(X, \mathcal{F}, P)$ is a measure space such that $m(X) = P(X) = 1$. P is a probability measure.

Ex 15.2: Any discrete space S is a probability space

where $\mathcal{F} = 2^S$, and $P(x_i) = p_i$ s.t. $\sum p_i = 1$.

c.g. a 52-card deck w/ $p_i = \frac{1}{52}$.

Ex 15.3: For $X = \mathbb{R}$, the normal distribution

$$P = \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$$

is a probability space for $\mathcal{M} = \mathcal{F}$ the Borel measurable sets.

$$P(E) = \int_E \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx.$$

More generally $X = \mathbb{R}^n$, $\mu \in \mathbb{R}^n$, Σ positive def $n \times n$

$$P = \frac{1}{(2\pi)^{n/2}} \frac{1}{\det \Sigma} e^{-\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)}.$$

15.1) Terminology Dictionary

Probability theory

Def 15.4: The following standard terminologies are equivalent

Measure theory (X, Σ, m)

Probability theory $(\Omega, \mathcal{F}, P)$

$x \in X$ point $\Rightarrow$

$x \sim P$ a sample
 $x \in \Omega$ ~~an event~~

$E \subseteq X$ measurable $\Rightarrow$

$E \subseteq \Omega$ an event

$m(E)$ measure $\Rightarrow$

$P(E)$ probability of E .

$f: X \rightarrow \mathbb{R}$ measurable function $\Rightarrow$

$X: \Omega \rightarrow \mathbb{R}$ random-variable

Ex 15.5: A random variable is an assignment of a number to each outcome. Eg $\Omega = \{P, N, D, Q\}$ coins,
 $V: \Omega \rightarrow \mathbb{R}$ value
 $\Omega = \text{cards}$
 $R: \Omega \rightarrow \mathbb{R}$ value ($J=11, Q=12 \dots$)

Note X_E is a R.V. Y_E measurable.

Def 15.6: The expectation value or average of a random variable is

$$IE(f) = \int_{\Omega} f dP$$

eg. $IE(f) = \sum_{x_i \in \Omega} f(x_i) p_i$ $IE(f) = \int_{\mathbb{R}} f(x) \frac{e^{-x^2/2}}{\sqrt{2\pi}} dx$.

Lemma 15.7: If (X, Σ, μ) is a ~~probability~~ ^{measure} space,
 $Y \in X$ measurable, then

$(Y, \{Y \cap E \mid E \in \Sigma\}, \mu(E \cap Y))$ is a measure space.

Proof: Obvious.

Def 15.8: If $(\Omega, \mathcal{F}, P)$ is a probability space, $E \in \mathcal{F}$ and event.
 $(E, \mathcal{F} \cap E, \tilde{P} = \frac{P(- \cap E)}{P(E)})$ is a probability space
 called the Conditional probability space.

$$P(A | E) = \tilde{P}(A) = \frac{P(A \cap E)}{P(E)} \quad \text{"the probability of A given E"}$$

the conditional probability

Corollary 15.9 (Bayes Rule) $P(A) P(A|E) = P(E \cap A) \cdot P(A)$.

Proof: Both are definitionally $P(A \cap E)$.

Rem 15.10: Radon-Nikodym theorem is used in defining conditional expectations.

15.ii) Some Inequalities

15.3

Thm 15.11 : The following hold

1) Suppose $X: \Omega \rightarrow \mathbb{R}$ is a positive R.V.

$$P(X \geq a) \leq \frac{E(X)}{a} \quad (\text{Markov Inequality})$$

2) X any R.V.

$$P(|X - E(X)| > a) \leq \frac{\text{Var}(X)}{a^2} \quad (\text{Chebyshev's Inequality})$$

$$\text{Var} = E(|X - E(X)|^2).$$

Proof : this is just translating notation.

15.iii) Large numbers and Central Limits

"The average of larger numbers of samples converges to the expectation value."

Def 15.12 : Let $\mathbb{I}: (X, \Sigma, \mu) \rightarrow (Y, \Sigma', -)$ a measurable mapping, the pushforward measure is defined by

$$\mathbb{I}_* \mu(B) = \mu(\mathbb{I}^{-1}(B)), \in \Sigma'.$$

! Pushforward of R.V. is a measure containing Borel sets.

Def 15.13 : Two random variables X_1, X_2 are said to be Identically distributed if

$$(X_1)_* P = (X_2)_* P. \quad \text{on } \mathbb{R}.$$

Def 15.14 : Two random variables X, Y are said to be independent if

$$P(X \leq a) \cdot P(Y \leq b) = P(X \leq a \text{ and } Y \leq b) \text{ on } \Omega \times \Omega.$$

i.e. those measurable functions on $\Omega \times \Omega$ of the form $f(x)g(y)$,

Thm 15.15 (Law of Large Numbers) Suppose that $X_1, X_2, \dots$ are a sequence of random variables that are identically distributed, and independent, w/

$$E(X_i) = \mu \quad \forall i.$$

Then for $\bar{X} = \frac{1}{N} \sum_{i=1}^N X_i$

1) (Weak Law)

$$\bar{X}_N \longrightarrow \mu \quad (\text{the constant function})$$

in probability (in measure)

2) (Strong Law)

$$\bar{X}_N \longrightarrow \mu$$

almost surely (pointwise a.e.).

Proof: Weak law assuming $\text{Var}(X_i) = \sigma^2 \quad \forall i$.

$$\begin{aligned} \text{Var}(\bar{X}_N) &= \text{Var}\left(\frac{1}{n}(X_1 + \dots + X_N)\right) \quad \checkmark \text{ Ind.} \\ &= \frac{1}{n^2} \text{Var}(X_1 + \dots + X_N) = \frac{1}{n^2} \sum \text{Var}(X_i) = \frac{\sigma^2}{n} \end{aligned}$$

Then

$$P(|\bar{X}_N - \mu| \geq \varepsilon) \leq \frac{\sigma^2}{n\varepsilon^2} \quad \text{by Chebyshev.}$$

$$\Rightarrow P(|\bar{X}_N - \mu| < \varepsilon) = 1 - \frac{\sigma^2}{n\varepsilon^2} \xrightarrow{n \rightarrow \infty} 1 \quad \forall \text{ fixed } \varepsilon. \quad \square$$

Thm 15.16 (Central Limit Theorem) Suppose X_i are independent, un $\mathbb{R}^d$ identically distributed, $E[X_i] = \mu$ and $\text{Var}(X_i) = \sigma^2 < \infty$.

Then

$$\sqrt{N}(\bar{X}_N - \mu) \longrightarrow \frac{1}{\sqrt{2\pi\sigma^2}} e^{-x^2/2\sigma^2} = N(0, \sigma^2)$$

in distribution $(P(\cdot \leq a) \longrightarrow P(N \leq a) \quad \forall a \in \mathbb{R})$.

Rem 15.17: By the way, $P(Y \leq a) = \int_{-\infty}^a g \, dx$ where g is the R-N density wrt Lebesgue measure.

Lecture 16 Banach Spaces I: L^p , and the fundamental inequalities. 16.1

Ex 16.1 : Consider the equivalence relation on E measurable,
 $f \sim g$ if $f = g$ a.e. $\Leftrightarrow \int_E |f - g| dx = 0$.

Def 16.2 : Let $(L^1(E), \|\cdot\|_1)$ denote the space of equivalence classes with the metric

$$d(f, g) = \|f - g\|_1 := \int_E |f - g| dx$$

Lemma 16.3 : $L^1(E)$ is a metric space.

Proof : $\|f - g\|_1 = 0$ iff $f \sim g$ so $f = g$ in L^1 by definition.

• $\|f - g\|_1 = \|g - f\|_1$ is clear.

• $\forall h$, by triangle inequality on $\mathbb{R}$,

$$\|f - g\|_1 = \int_E |f - g| \leq \int_E |f - h| + |h - g| \leq \int_E |f - h| + \int_E |h - g| \leq \|f - h\|_1 + \|h - g\|_1$$

Prop 16.4 : $L^1(E)$ is complete, i.e., if $f_n \in L^1(E)$ is Cauchy, then $f_n \rightarrow f$ in $L^1(E)$ and $\|f_n - f\|_1 \rightarrow 0$.

Proof : Note that if $|g_n| \geq 0$, $\sum_{n=1}^{\infty} \int |g_n| < \infty$ then monotone convergence implies (apply to $G_N = \sum_{n=1}^N |g_n|$)

$$\sum_{n=1}^{\infty} \int |g_n| = \lim_{N \rightarrow \infty} \int G_N = \int \sum_{n=1}^{\infty} |g_n|. \quad \text{In particular, } \sum_{n=1}^{\infty} |g_n| \text{ is integrable.}$$

Now let f_{n_k} be a subsequence so $\|f_{n_k} - f_{n_{k+1}}\|_1 \leq \frac{1}{2^k}$.

$$f = f_{n_1} + \sum_{k=1}^{\infty} f_{n_{k+1}} - f_{n_k}$$

$$g = |f_{n_1}| + \sum_{k=1}^{\infty} |f_{n_{k+1}} - f_{n_k}|$$

By the above, $\int g < \infty$, since $\sum_{k=1}^{\infty} \frac{1}{2^k} < \infty$, so $f \in L^1$ integrable. Then claim.
 $f_{n_k} \rightarrow f$ a.e.

$$m(|f_{n_k} - f_{n_k}| > \frac{1}{\sqrt{2^k}}) \leq \frac{1}{\sqrt{2^k}} \int |f_{n_{k+1}} - f_{n_k}| \leq \frac{1}{\sqrt{2^k}} \text{ by Markov.}$$

so $\sum m(|f_{n_{k+1}} - f_{n_k}|) < \infty \Rightarrow m(x | x \text{ in infinitely many}) = 0$
 $m(x | f_{n_k} \text{ doesn't converge}) = 0$
 by Borel - Cantelli.

Then $f_{n_k} \rightarrow f$ in L^1 by dominated convergence. Now choose N large
 $\|f_n - f\| \leq \|f_{n_k} - f_n\| + \|f_{n_k} - f\| < \frac{\epsilon}{2} + \frac{\epsilon}{2} \quad \blacksquare$

16.ii) Banach Spaces

Def 16.5: A norm on a vector space is a function
 $\|\cdot\|: X \rightarrow \mathbb{R}^{\geq 0}$
 such that $(X, \|\cdot\|)$ is a normed vector space.

- i) $\|\alpha x\| = |\alpha| \|x\|$
 - ii) $\|x\| = 0 \iff x = 0$
 - iii) $\|x+y\| \leq \|x\| + \|y\|$.
- Note $d_X(x,y) = \|x-y\|$ is therefore a metric.

Def 16.6: A normed vector space is a Banach Space if
 X is complete as a metric space.

Ex 16.7: $\ell^1 \cong \mathbb{R}^{\infty}$ is the space of sequences $a = (a_0, a_1, \dots)$
 with $\|a\|_{\ell^1} = \sum_{i=1}^{\infty} |a_i|$
 completeness follows from completeness of $\mathbb{R}$.

Ex 16.8 $\ell^2 = \{a_0, a_1, \dots\}$
 $\|a\|_{\ell^2} = \left(\sum_{i=1}^{\infty} |a_i|^2 \right)^{1/2}$
 $\|a\|_{\ell^p} = \left(\sum_{i=1}^{\infty} |a_i|^p \right)^{1/p}$
 $\|a\|_{\ell^\infty} = \sup_i |a_i|$
 (all complete, different spaces)

Lemma 16.1 : If $Y \subseteq X$ is a vector subspace closed in the induced topology then Y is also a Banach space. 16.3

Ex 16.10 : Let $K \subseteq \mathbb{R}$ be compact, then

$C^0(K) := \{f \mid f \text{ continuous, } \|f - g\|_\infty = \sup_x |f(x) - g(x)|\}$
is a Banach space.

$C^\alpha(K) = \{f \mid f \text{ Hölder continuous, } 0 < \alpha \leq 1\}$
 $\|f\|_{C^\alpha} = \sup |f(x)| + \sup_{x,y} \frac{|f(x) - f(y)|}{|x - y|^\alpha}$
 $\alpha = 1$ is Lipschitz continuous.

Def 16.11 : Let E be a measurable set then the L^p spaces

$$L^p(E) = \{f: E \rightarrow \mathbb{R} \mid f \text{ measurable, } \|f\|_{L^p} < \infty\}$$

where $\|f\|_{L^p} = \left(\int_E |f|^p dx \right)^{1/p}$, is the L^p -norm

For $1 \leq p < \infty$, and

$$\|f\|_{L^\infty} = \inf \{M \mid |f| \leq M \text{ a.e.}\}.$$

16.iii) Minkowski, Young, and Hölder Inequalities.

Thm 16.12 (Minkowski) If $1 \leq p < \infty$, $\|f + g\|_{L^p} \leq \|f\|_{L^p} + \|g\|_{L^p}$
w/ equality iff $f = \alpha g$.

Proof : It suffices to show the unit ball is convex, ~~le~~ by linearity
may assume $\|f\| + \|g\| = 1$. Then $\bar{f} = \frac{f}{\|f\|_{L^p}}$, $\bar{g} = \frac{g}{\|g\|_{L^p}}$

so $f + g = \alpha \bar{f} + \beta \bar{g}$. $\bar{f}, \bar{g} \in B$, and $\alpha + \beta = 1$,

so can show $\|f + g\| = \|\alpha \bar{f} + \beta \bar{g}\| \leq 1 = \|f\| + \|g\|$
 $\downarrow$
 $\in B$.

Thus $\forall t \in [0, 1]$ need $\|tf + (1-t)g\|_{L^p} \leq 1$. or $\|tf + (1-t)g\|_{L^p}^p \leq 1$

$$\int |tf + (1-t)g|^p \leq \int t|f|^p + (1-t)|g|^p \leq 1 \quad (x^p \text{ is convex})$$

Thm 16.13 (Riesz-Fischer) For $1 \leq p \leq \infty$, L^p is complete, and 16.4
a Banach space.

Proof: The triangle inequality is precisely Minkowski's inequality.

For $1 < p < \infty$, the proof mimics $p=1$.

Suppose f_n is Cauchy, set $\|f_{n_k} - f_{n_{k+1}}\| \leq \frac{1}{2^k}$

$$f = f_{n_1} + (f_{n_2} - f_{n_1}) + \dots$$

$$g = |f_{n_1}| + |f_{n_2} - f_{n_1}| + \dots$$

then $\lim_{n \rightarrow \infty} \|G_n\|_{L^p} = \left(\int \left(\sum |g_n| \right)^p \right)^{1/p}$

$$\leq \left(\int \left(\sum |g_n|^p \right)^{1/p} \right)^{1/p}$$

~~convex~~
Minkowski obvious for $p = \infty$.
(apply Minkowski N times)

$$\leq \sum_{n=1}^N \left(\int |g_n|^p \right)^{1/p}$$

$$\leq \infty.$$

so ~~$f_n \in L^p$~~ $g \in L^p \Rightarrow f \in L^p$. Rest same w/ Chebychev, and $|g|^p$ in dominated convergence. $\square$

Lemma 16.14: If $m(A) < \infty$ and $f \in L^p, L^\infty$ for all p ,

$$\lim_{p \rightarrow \infty} \|f\|_{L^p} = \|f\|_{L^\infty}$$

Proof: If χ is a step function

$$\|\chi\|_{L^p} = m(A)^{1/p} \rightarrow 1 = \|\chi\|_{L^\infty}$$

Now take limits. $\square$

Lecture 17 Banach Spaces II : ~~Density~~. Density, Young-Hölder 17.18

Def 17.1 : A Banach space has a dense subset (subspace)

$$S \subseteq X$$

if $\forall x \in X, \varepsilon > 0 \exists s \in S$ with $\|s - x\| < \varepsilon$.

X is separable if $\exists$ a countable, dense subset.

Prop 17.2 : For $1 \leq p < \infty$ simple functions are dense.

For $1 \leq p < \infty$ step, continuous, smooth functions are dense.

Proof : For L^∞ , take f , so $M = \|f\|_{L^\infty}$. Divide $[-M, M]$ up and take $\sum_{n=1}^N (M - \frac{n}{N}) \cdot \chi_{f^{-1}[(M - \frac{n}{N}, M - \frac{n-1}{N}]}$

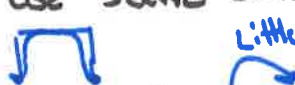
For $p \neq \infty$, let $f_M = \min_{\text{if } \infty} [M, f] \cdot \chi_{[-M, M]}$.

Since $\|f_M - f\| \rightarrow 0$ pointwise, dominated by $|f|^p$

$$\int |f_M - f|^p \rightarrow 0.$$

If $S \subseteq T \subseteq X$ are dense inclusions, $S \subseteq X$ is dense.

For f_M use same trick as L^∞ (since $f_M \in L^1 \cap L^p \cap L^\infty$ and compact support)

 smooth $\subseteq$ step $\subseteq$ simple.

Prop 17.3 : L^p is separable for $p \neq \infty$.

Proof : Step functions $\sum_{i=1}^N q_i \chi_{[p_i, t_i]}$ $q_i, p_i, t_i \in \mathbb{Q}$
are dense in step functions.

17.ii Young and Hölder Inequalities

Thm 17.4 : Suppose $\frac{1}{p} + \frac{1}{q} = 1$. Then

$$\|fg\|_{L^1(E)} \leq \frac{\|f\|_{L^p(E)}^p}{p} + \frac{\|g\|_{L^q}^q}{q}.$$

Proof: It suffices to show

$$ab \leq \frac{a^p}{p} + \frac{b^q}{q} \quad (\text{note } p=q=2 \text{ is } 2|ab| \leq a^2 + b^2 \text{ is clear})$$

Since $\ln$ is concave, $t = \frac{1}{p}$ $1-t = \frac{1}{q}$

$$\begin{aligned} \ln(ta^p + (1-t)b^q) &\geq \frac{1}{p} \ln(a^p) + \frac{1}{q} \ln(b^q) \\ &= \ln(a) + \ln(b) = \ln(ab). \end{aligned}$$

Take exp.

Corollary 17.5 (Young w/ ε) or "Peter-Paul" or "Absorption"

$\forall \varepsilon > 0$

$$\|fg\|_{L^1} \leq \frac{\varepsilon^p \|f\|_{L^p}^p}{p} + \frac{\|f\|_{L^q}^q}{q \varepsilon^q}$$

Proof: Apply Young to $fg = \varepsilon f \cdot \frac{g}{\varepsilon}$.

Rem 17.6: The above is, in my opinion, the most useful fact in analysis.

Thm 17.7 (Hölder's Inequality) Suppose $\frac{1}{p} + \frac{1}{q} = 1$. Then (including $p=q=\infty$)

$$\|fg\|_{L^1} \leq \|f\|_{L^p} \|g\|_{L^q}$$

Proof: Set $\bar{f} = \frac{f}{\|f\|_{L^p}}$ $\bar{g} = \frac{g}{\|g\|_{L^q}}$.

$$\text{Then } \|\bar{f}\bar{g}\|_{L^1} \leq \frac{\|\bar{f}\|_{L^p}^p}{p} + \frac{\|\bar{g}\|_{L^q}^q}{q} = \frac{1}{p} + \frac{1}{q} = 1. \text{ By Young.}$$

$$\Rightarrow \|fg\| \leq \|f\|_{L^p} \|g\|_{L^q}$$

$$\text{If } q = \infty, \quad \int |fg| \leq \sup f \cdot \int |g|.$$

□

17.iii) Examples17.3Ex 17.8 : Suppose $m(E)$ is finite.

Then $q \geq p$:

$$\int |f|^q \leq \left(\int |f|^{p \cdot \frac{q}{p}} \right)^{1/q} \left(\int 1^{q \cdot \frac{p}{p}} \right)^{1/p}$$

$$\frac{1}{q} + \frac{1}{q/p} = 1.$$

$$\leq \|f\|_{L^p} m(E)^{1/p}$$

so finite L^q implies finite L^p so $L^q \subseteq L^p$.

Thus

~~$L^1(E) \subseteq L^2(E) \subseteq L^p(E) \subseteq L^q(E) \subseteq L^\infty(E)$~~

are strict inclusions.

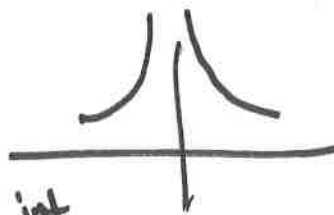
Intuition 17.1 : Large p makes mass escaping vertically more problematic.
Small p horizontally.

Ex 17.10 : $E = \mathbb{R}$. $p \leq q$. Then $\exists f \in L^p \setminus L^q$ and vice versa.

Take

$$f = \frac{1}{x^{1/q}} \chi_{[-1,1]}.$$

$$|f|^q = \frac{1}{x} \text{ not int, } |f|^p = \frac{1}{x^{p/q}} \text{ int.}$$



Take

$$f = \frac{1}{x^{1/p}} \chi_{[1,\infty)}$$

$$|f|^q = \frac{1}{x^{q/p}} \text{ integrable}$$

$$|f|^p = \frac{1}{x} \text{ not integrable.}$$



Rem 17.11: Recall every metric space has a completion $X \hookrightarrow \bar{X} = \{\text{equiv class of Cauchy sequences}\}$ 17.4

The density shows that

$(C^0[a,b], \| \cdot \|_{L^p}) \hookrightarrow L^p([a,b])$
is this completion. So all we've done in some sense is very precisely characterize this completion.

17.iv) Dual Spaces

Def 17.12

A linear functional $\varphi: X \rightarrow \mathbb{R}$ is said to be bounded if

$$|\varphi(x)| \leq C \|x\| \quad \forall x \in X.$$

Def 17.13: Suppose X is a Banach space, then

$$X^* = \{ \varphi \mid \varphi: X \rightarrow \mathbb{R} \text{ bounded} \}, \quad \|\varphi\|_{X^*} = \sup_{\|x\|=1} |\varphi(x)|.$$

Prop 17.14: X^* is a Banach space.

Proof: Linear functionals are linear, and triangle inequality on $\mathbb{R}$ shows norm.

Suppose φ_i is Cauchy, so that

$$\|\varphi_i - \varphi_j\|_{X^*} = \sup_{\|x\|=1} |\varphi_i(x) - \varphi_j(x)| \rightarrow 0.$$

Then $\varphi_i(x)$ is Cauchy $\forall x$, set $\varphi(x) = \lim_{i \rightarrow \infty} \varphi_i(x)$. It is a simple matter to show $\varphi(x)$ is linear. $\square$

Lemma 17.15: There is a natural inclusion $X \hookrightarrow X^{**}$.

If it is an isomorphism, X is called Reflexive.

Proof: Elements of $X^{**} = \{ \xi: X^* \rightarrow \mathbb{R} \mid \text{bounded linear} \}$.

$x \mapsto \xi_x$ defined by $\xi_x(\varphi) := \varphi(x)$.
Then obviously linear,
 $\|\xi_x\| \leq \|\varphi\| \|x\|$ so bounded. $\square$

Lecture 18 Banach Spaces III: Riesz Representation and Weak Topology 18.1

Recall each Banach space has a dual

$$X^* = \{ \varphi \mid \|\varphi(x)\|_X \leq \|\varphi\|_{X^*} \|x\| < \infty \}$$

Question 18.1: What is the dual of $L^p(E)$.

Ex 18.2: There is an inclusion $L^q(E) \hookrightarrow (L^p(E))^*$ for $\frac{1}{p} + \frac{1}{q} = 1$.

$$f \longmapsto \underbrace{\int_E f \, dx}_{\varphi_f}$$

It is bounded because by Hölder,

$$\|\varphi_f(g)\|_{\mathbb{R}} \leq \|fg\|_{L^1} \leq \|f\|_{L^q} \|g\|_p$$

$$\text{So } \|\varphi_f\|_{(L^p)^*} = \|f\|_{L^q}.$$

Theorem 18.3 (Riesz Representation) If $\frac{1}{p} + \frac{1}{q} = 1$ w/ $1 < p, q < \infty$,

then

$$L^q(E) \xrightarrow{\sim} (L^p(E))^*$$

$$f \longmapsto \varphi_f$$

is an isometry.

! We will prove for $E = \mathbb{R}$ or $[a, b]$

Corollary 18.4: L^p is reflexive.

Proof: Given $\varphi \in (L^p(E))^*$ define a candidate

$$F(x) = \begin{cases} \varphi(\chi_{[0, x]}) & x > 0 \\ -\varphi(\chi_{[x, 0]}) & x < 0 \end{cases}$$

Claim F is absolutely cont.

$$\begin{aligned} \sum_{i=1}^N |F(b_i) - F(a_i)| &= \left| \sum_{i=1}^N \varphi(\chi_{[a_i, b_i]}) \right| \\ &\leq \varphi\left(\sum_{i=1}^N \chi_{[a_i, b_i]}\right) \leq C_\varphi \left\| \sum_{i=1}^N \chi_{[a_i, b_i]} \right\|_{L^p} \\ &\leq C_\varphi \delta^{1/p}. \end{aligned}$$

Thus $\forall [n, n+1] \exists f$ st $f(x) = F'(x)$ a.e.

18.2

Then

$$\varphi_f(\chi_{[a,b]}) = \int f \chi_{[a,b]} dx = \int_a^b f dx = F(b) - F(a) = \varphi(\chi_{[a,b]})$$

agree on step functions + continuity / density.

It remains to show $f \in L^q(\mathbb{R})$. Suppose first $f \geq 0$.

Let $f_m = \min(m, f) \chi_{[-m, m]}$.

$$\int_{\mathbb{R}} |f_m|^q \leq \int_{\mathbb{R}} |f_m|^{q-1} |f_m| = \varphi_f(|f_m|^{q-1})$$

$$\begin{aligned} &\leq C_\varphi \|f_m^{q-1}\|_{L^p} \\ &\leq C_\varphi \left(\int |f_m|^{(q-1)p} \right)^{1/p} \\ &\leq C_\varphi \left(\int |f_m|^q \right)^{1/p} \end{aligned}$$

$$\frac{1}{p} + \frac{1}{q} = 1 \Rightarrow p+q = pq$$

$$\text{so } pq - p = q$$

$$\frac{q}{q(1-\frac{1}{p})} = q \cdot \frac{1}{1-\frac{1}{p}} = 1$$

$$\text{so } \|f_m\|_{L^q} \leq C_\varphi$$

Then $M \rightarrow \infty$ by monotone convergence, $\|f\|_{L^q} \leq C_\varphi$.

For f general use $f = f^+ \chi^+ - f^- \chi^-$ trick.

Thm 18.5 (Reisz Rep $p = \infty$) $L^1(\mathbb{R})^* = L^\infty(\mathbb{R})$, but not vice versa.

Proof: $F(x)$ as before,

~~extend φ to L^∞~~ extend φ .

$$\|F(b) - F(a)\|_{\mathbb{R}} \leq \|\varphi\|_* \|\chi_{[a,b]}\|_{L^1} = (b-a) \|\varphi\|_*$$

so absolutely cont as before,

$$f = F' \text{ a.e.}$$

$$\text{and } \|f\|_{L^\infty} \leq \|\varphi\|_*$$

$$\text{Again } \varphi_f(\chi_{[a,b]}) = \int_a^b F' = F(b) - F(a) = \varphi(\chi_{[a,b]})$$

agree, so equal by density.

18.4) The Weak Topologies

18.3

Def 18.6 a sequence $f_n \in X$ to converge weakly to f
 $f_n \rightarrow f$

if $\varphi(f_n) \rightarrow \varphi(f)$ in $\mathbb{R} \quad \forall \varphi \in X^*$.

Ex 18.7: $f_n = \chi_{[n, n+1]} \rightarrow 0$ in $L^2(\mathbb{R})$ but not strongly. By Riesz $(L^2)^* = L^2$

$$\text{So } \int_{\mathbb{R}} \chi_{[n, n+1]} g \leq \|g\|_{L^2[n, n+1]}$$

$$\text{but } \sum_{n=1}^{\infty} \|g\|_{L^2[n, n+1]} < \infty \text{ so } \rightarrow 0.$$

$$\text{but } \|f_n\|_{L^2} = 1 \quad \forall n.$$

Prop 18.8: If $f_n \rightarrow f$ in $L^p(E)$ $1 \leq p < \infty$
 then $\exists M$ st $\|f_n\|_{L^p} < M$.

Proof: Suppose not, By passing to a subsequence,
 May assume $f = 0$. $\|f_n\| \geq n \cdot 3^n$.

Renormalize so

$$g_n = \frac{n \cdot 3^n}{\|f_n\|} \cdot f$$

bounded by $[0, 1]$
 so sub converges.

$$\text{Claim } \|g_n\| = n \cdot 3^n \text{ and } g_n \rightarrow g \text{ weakly } \varphi(g_n) = \frac{n \cdot 3^n}{\|f_n\|} \varphi(f_n)$$

$$\text{Now let } \varepsilon_1 = \frac{1}{3}$$

$$\varepsilon_{n+1} = \frac{1}{3^{n+1}} \cdot \text{sign} \left(\int_E \left[\sum_{k=1}^n \varepsilon_k (f_k)^* \right] f_n \right)$$

$$f_n^* = |f_n|^{\frac{p}{p-1}} \text{sgn}(p) |f_n|^{p-1}$$

$$\in L^q \quad \left(\int f_n^* f_n \right)^{1/p} = \|f_n\|_p$$

$$\text{so } \left| \int_E \sum_{k=1}^n \varepsilon_k f_k^* f_n \right| \geq \frac{1}{3^n} \|f_n\|_p = n$$

$$\|f_n^*\|_q = 1 \text{ so } \|\varepsilon_n f_n^*\|_q = \frac{1}{3^n}$$

Thus $\sum_{k=1}^{\infty} \varepsilon_k f_k^* \rightarrow g$ in L^q .

(18.4)

$$\begin{aligned} |\int g \cdot f_n| &\geq |\int \sum_{k=1}^{\infty} \varepsilon_k f_k^* f_n| - \int \sum_{k=n+1}^{\infty} \varepsilon_k f_k^* \\ &\geq n - \|\sum_{k=n+1}^{\infty} \varepsilon_k f_k^*\|_q \|f_n\|_p \\ &\geq n - \frac{1}{3^n} \|f_n\|_p \rightarrow 0. \quad \square \end{aligned}$$

Thm 18.9: Let $1 < p < \infty$. Then a sequence $\{f_n\}$ (Banach-Alaoglu) has a weakly convergent subseq if it has a bounded subsequence.

i.e. "The unit ball is weakly compact".

Proof: See Royden 8.3.

Rem 18.10: This is a great result. If you want to show $f_k \rightarrow f$ (say the minimum of a functional or sol of PDE)

Then can split as two easier results

- 1) f_k is bounded $\Rightarrow \exists f$ st $f_k \rightarrow f$.
- 2) convergence is actually strong (no escaping mass).

Lecture 19 Fourier Series I: Hilbert spaces.

Def 19.1: A linear operator $L: X \rightarrow Y$ on Banach spaces is said to be bounded if $\exists C = C_L$ such that

$$\|Lx\|_Y \leq C_L \|x\|_X.$$

Lemma 19.2: A linear operator is continuous iff it is bounded.

Proof: $\forall \varepsilon > 0$ take $\delta = \frac{\varepsilon}{C_L}$ so

$$\|Lx - Ly\|_Y = \|L(x-y)\|_Y \leq C_L \|x-y\|_X \leq \varepsilon \quad \text{if } \|x-y\|_X < \delta.$$

Take $\varepsilon = 1$. $\exists \delta$ s.t. if $\|x-0\|_X < \delta$ then for $\|x\|_X = \delta$

$$\|Lx\|_Y \leq 1 \leq \frac{1}{\delta} \|x\|_X, \quad \text{and both sides scale.}$$

Def 19.2: Two Banach spaces X_1, X_2 are isomorphic if $\exists$ a bounded linear isomorphism w/ bounded inverse.

19.i) Hilbert Spaces

Def 19.3: An inner product space H with

$$\langle \cdot, \cdot \rangle: H \times H \rightarrow \mathbb{R}$$

bilinear is a Hilbert space if the norm $\|h\|_H = \sqrt{\langle h, h \rangle}$ makes it into a Banach space.

Ex 19.4: $L^2(\mathbb{R})$ is a Hilbert space with

$$\langle f, g \rangle_{L^2} = \int_{\mathbb{R}} fg \, dx$$

Hölder's inequality specializes to Cauchy-Schwartz

$$|\langle f, g \rangle_{L^2}| \leq \|f\|_{L^2} \|g\|_{L^2}$$

Rem 19.5: In the case of a general inner product space Cauchy-Schwartz follows from general nonsense.

Def 19.6: A collection of vectors $\{e_i\}$ is said to be an orthonormal set if $\langle e_i, e_j \rangle = \delta_{ij}$.

Def 19.7: An orthonormal set is said to be a basis if it is complete, i.e. the equivalent conditions hold

- 1) If $\langle x, e_i \rangle = 0 \forall i$, then $x = 0$.
- 2) $\sum_{i=1}^{\infty} a_i e_i$ finite linear combinations are dense.

Proof: Suppose dense. $\|x - \sum_{i=1}^N a_i e_i\| < \varepsilon$, but $0 = \langle x, \sum_{i=1}^N a_i e_i \rangle$
 $= \langle x, \sum_{i=1}^N a_i e_i - x \rangle + \langle x, x \rangle$
 $= \|x\|^2 - \varepsilon \|x\| \rightarrow \leftarrow$

Suppose complete, if not dense $\exists y$ w/
 $\|y - \sum_{i=1}^N a_i e_i\| > \delta \quad \forall a_i \in \mathbb{R}$ for some δ .

Then $y - \sum_{i=1}^{\infty} \langle y, e_i \rangle e_i$ has $\langle y, e_i \rangle = 0 \quad \forall i \rightarrow \leftarrow$

Algorithm 19.8 (Gram-Schmidt)

Let a_i be a dense subset of H , for $i=1, 2, \dots$

Set $e_1 = a_1$
 $\tilde{e}_{n+1} = a_{n+1} - \sum_{i=1}^n \langle a_{n+1}, e_i \rangle e_i$ then $e_i = \frac{\tilde{e}_i}{\|\tilde{e}_i\|}$.

Corollary 19.8: A separable Hilbert space has a (countable) orthonormal basis, hence

$$x = \sum_{i=1}^{\infty} \langle x, e_i \rangle e_i \quad \forall x \in H.$$

Proof: Existence follows from Gram-Schmidt on dense subspace

Note $a_i = \langle x, e_i \rangle$ has $\sum |a_i|^2 \leq \|x\|^2$

Then

$$0 \leq \|x - \sum_{i=1}^N a_i e_i\|^2 = \|x\|^2 + \sum_{i=1}^N |a_i|^2 - 2 \langle x, \sum_{i=1}^N a_i e_i \rangle$$

$$= \|x\|^2 - \sum_{i=1}^N |a_i|^2$$

so $\sum_{i=1}^{\infty} \langle x, e_i \rangle e_i$ converges. By completeness, $x - \text{this} = 0$.

Corollary 19.10: $\exists$ an isomorphism $H \rightarrow \ell^2(\mathbb{N})$ for each basis (in fact an isometry)

Proof: $x \rightarrow \{\langle x, e_i \rangle\}_{i=1}^{\infty}$

Thm 19.11 (Abstract Riesz Representation) Let H be a separable Hilbert space, then $H \rightarrow H^*$
 $x \mapsto \langle x, - \rangle$
 is an isomorphism. 19.3

Proof: Obviously linear, and $\sup_{\|f\|=1} \langle x, f \rangle = \|x\|$ so bounded, and obviously inj.

Let $\varphi \in H^*$. Set $a_i = \varphi(e_i)$. Then

$(\varphi - \langle \sum a_i e_i, - \rangle) e_j = \varphi(e_j) - a_j = 0$
 so $\varphi = \sum_{i=1}^{\infty} a_i e_i$, hence surj. (isometry so inverse is bounded).

19.ii) Examples

Ex 19.12: $\ell^2(\mathbb{N})$ or $\ell^2(\mathbb{Z})$ with bases $e_i = \{0, 0, 0, \dots, 1, 0, \dots\}$

Ex 19.13: The quantum mechanics Hilbert space is (usually) $L^2(\mathbb{R}^3)$.

Ex 19.14: More generally (X, Σ, μ) any measure space $L^2(X; \mu)$.

Ex 19.15 $L^2[-1, 1]$ a basis is the Legendre polynomials
 $1, x, \frac{1}{2}(3x^2-1), \frac{1}{2}(5x^3-3x), \frac{1}{8}(35x^4-30x^2+3), \dots$
 (Gram-Schmidt on x^n), density below. $= \frac{1}{2^n n!} \left(\frac{d}{dx} \right)^n (x^2-1)^n$

Ex 19.16: $L^2(\mathbb{R}, e^{-x^2/2})$

$$h_n(x) = (-1)^n e^{x^2/2} \left(\frac{d}{dx} \right)^n e^{-x^2/2} = (2x - \frac{d}{dx})^n \cdot 1.$$

(Hermite Polynomials)

$$= 1, 2x, 4x^2-2, 8x^3-12x, \dots$$

are orthonormal basis.

19.iii) Fourier Series

19.4

Def 19.17: an algebra of functions on K compact

- 1) separates points if $\forall x, y \in K, \exists f$ w/ $f(x) \neq f(y)$
- 2) vanishes nowhere if $\nexists x \in K$ w/ $f(x) = 0 \forall f$.

Thm 19.18 (Stone-Weierstrass) Any algebra $\mathcal{F} \subseteq C^0(K)$ that separates points and vanishes nowhere is dense in $C^0(K)$.

Upgrade 19.19: Consider $\mathbb{C}$ -valued functions.

$L^2([a, b])$ works same w/ $\langle f, g \rangle = \int_a^b f \bar{g} dx$

Also $L^2([a, b]) \cong L^2(S^1)$ as completion of $f(a) = f(b)$ in C^0 .

⇒ For next few lectures we work with $L^2(S^1; \mathbb{C})$, $\langle f, g \rangle = \frac{1}{2\pi} \int_0^{2\pi} f \bar{g} d\theta$.

Thm 19.20: $\{e^{in\theta}\}_{n=-\infty}^{\infty}$ for $\theta \in [0, 2\pi]$ is an orthonormal basis for $L^2(S^1; \mathbb{C})$.

Proof: $\therefore \frac{1}{2\pi} \int_0^{2\pi} |e^{in\theta}|^2 d\theta = \frac{1}{2\pi} \cdot 2\pi = 1$

$$\cdot \frac{1}{2\pi} \int_0^{2\pi} \langle e^{in\theta}, e^{im\theta} \rangle d\theta = \frac{1}{2\pi} \int_0^{2\pi} e^{i(n-m)\theta} d\theta = \frac{1}{i(n-m)} e^{i(n-m)\theta} \Big|_0^{2\pi} = 0$$

• $\operatorname{Re}(e^{in\theta})$ and $\operatorname{Im}(e^{in\theta})$ never simultaneously vanish

• ~~Since $\operatorname{const} \in C^0$, if $\nexists x, y$ then $\nexists x, y$ st $f(x) \neq f(y) \neq 0$~~

$e^{i\theta}: S^1 \rightarrow S^1 \subseteq \mathbb{C}$ is already injective.

Def 19.21: The isomorphism/isometry

$$\mathcal{F}: L^2(S^1; \mathbb{C}) \longrightarrow L^2(\mathbb{Z}; \mathbb{C})$$

$$f \longmapsto \langle e^{in\theta}, f \rangle$$

is called the Fourier Series, and denoted

$$\mathcal{F}(f) = \hat{f}(n): \mathbb{Z} \rightarrow \mathbb{C}.$$

Lecture 20 Fourier Series II: Derivatives and convolutions

20.1

Recall Fourier series gave an isomorphism

$$L^2(S^1; \mathbb{C}) \cong \ell^2(\mathbb{Z}; \mathbb{C})$$

$$f \mapsto \hat{f}(n) = \langle f, e^{in\theta} \rangle_{L^2}$$

$$\sum_{n=-\infty}^{\infty} c_n e^{in\theta} \longleftrightarrow \{c_n\}$$

In fact
Thm 20.1 (Plancherel) $\mathcal{F}$ is an isometry,

$$\|f\|_{L^2} = \sum_{n \in \mathbb{Z}} |c_n|^2 \quad \hat{f}(n) = c_n.$$

Ex 20.2 : For $c_n = a_n + ib_n$ $c_n e^{in\theta} = (a_n + ib_n)(\cos(n\theta) + i\sin(n\theta))$

(Real Fourier Series)

And

$$= a_n \cos(n\theta) - b_n \sin(n\theta) + i(b_n \cos(n\theta) + a_n \sin(n\theta)).$$

$$\langle f + ig, \cos(n\theta) + i\sin(n\theta) \rangle_{\mathbb{C}} = \langle f, \cos \rangle + \langle g, \sin \rangle + i \langle f, -\sin \rangle + i \langle g, \cos \rangle.$$

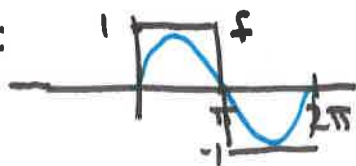
$\Rightarrow$ if f is real,

$$f = \operatorname{Re} \left(\sum_n c_n e^{in\theta} \right) = \sum_n a_n \cos(n\theta) + b_n \sin(n\theta)$$

$$\text{where } a_n = \langle f, \cos(n\theta) \rangle_{L^2}$$

$$b_n = \langle f, -\sin(n\theta) \rangle_{L^2}$$

Ex 20.3 :



then $a_n = 0 \quad \forall n$ b/c function is odd.

$$-b_n = \frac{1}{2\pi} \int_0^{2\pi} f(x) \sin(nx) dx$$

$$= \frac{1}{\pi} \int_0^{\pi} \sin(nx) dx$$

$$= \frac{1}{\pi n} [-\cos(nx)]_0^{\pi} = \begin{cases} \frac{2}{\pi n} & n \text{ even} \\ 0 & n \text{ odd} \end{cases}$$

$$f = \sum_{n \text{ odd}} (-1)^{\frac{n-1}{2}} \frac{2}{\pi n} \sin(nx).$$

$$= \sum_{n=1,3,5} \frac{4}{\pi n} \sin(nx).$$

20.i) The Derivative

20.2

Observe $-i\partial_\theta e^{in\theta} = n e^{in\theta}$.

Prop 20.4 : The Fourier series satisfies (for $f \in C^\infty$)

$$\mathcal{F}(-i\partial_\theta f) = n \hat{f}(n)$$

Proof : Let $g = -i\partial_\theta f \in L^2$.

$$g = \sum_{n=-\infty}^{\infty} d_n e^{in\theta} \quad \text{where}$$

$$f = \sum_{n=-\infty}^{\infty} c_n e^{in\theta} \quad \text{where} \quad c_n = \frac{1}{2\pi} \int_0^{2\pi} \langle f, e^{in\theta} \rangle d\theta$$

$$\begin{aligned} \text{And} \quad d_n &= \frac{1}{2\pi} \int_0^{2\pi} \langle -i\partial_\theta f, e^{in\theta} \rangle d\theta = \frac{1}{2\pi} \int_0^{2\pi} \langle \partial_\theta f, i e^{in\theta} \rangle d\theta \\ &= \frac{1}{2\pi} \int_0^{2\pi} \langle f, -i\partial_\theta e^{in\theta} \rangle d\theta \\ &= n \cdot c_n \end{aligned}$$

$$\text{Hence} \quad \sum_{n=-\infty}^{\infty} |n|^2 |c_n|^2 < \infty, \quad \text{and} \quad \left\| \sum_{n=-N}^N n c_n e^{in\theta} - (-i\partial_\theta f) \right\|_{L^2} \xrightarrow{N \rightarrow \infty} 0.$$

Corollary 20.5 : $(-i\partial_\theta)^k f \in L^2$ if and only if $\sum_{n=-\infty}^{\infty} |n|^{2k} |c_n|^2 < \infty$.

Proof : $\Rightarrow$ induction on above.

$$\Leftarrow \lim_{N \rightarrow \infty} \int_0^{2\pi} \left| \frac{f_N(\theta+h) - f_N(\theta)}{h} \right|^2 d\theta$$

$-i\partial_\theta f_N \rightarrow g$ in L^2 . Claim $g = -i\partial_\theta f$. exchange limits + dominated convergence.

20.ii) The Convolution

Question 20.6: What is $\mathcal{F}(fg)$?

Thm 20.7 : If $f, g \in L^2$ with $g \in \mathcal{O}'(S; \mathbb{C})$. Then

$$\begin{aligned} \mathcal{F}(fg) &= \hat{f} * \hat{g} \\ &= \sum_{n \in \mathbb{Z}} \sum_{k \in \mathbb{Z}} \hat{f}(n-k) \hat{g}(k) \end{aligned}$$

Proof : First note
 $(\hat{f} * \hat{g}) \in \ell^2$, $\therefore$

20.3

$$\begin{aligned} \|\hat{f} * \hat{g}\|_{\ell^2} &= \sum_n \left| \sum_{k \in \mathbb{Z}} \hat{f}(n-k) \hat{g}(k) \right|^2 \\ &\leq \sum_n \left(\left(\sum_k \left| \frac{\hat{f}(n-k)}{(k^2+1)} \right|^2 \right)^{1/2} \left(\sum_k (k^2+1) |\hat{g}(k)|^2 \right)^{1/2} \right)^2 \\ &\leq \sum_n (\|\hat{g}\|_{L^2} + \|d\hat{g}\|_{L^2}) \sum_k \frac{|\hat{f}(n-k)|^2}{(k^2+1)^2} \\ &\leq \sum_k \sum_n \frac{|\hat{f}(n-k)|^2}{(k^2+1)^2} \leq (\|\hat{g}\|_{L^2} + \|d\hat{g}\|_{L^2}) (\|f\|_{L^2}) \sum_k \frac{1}{k^4} \\ &< \infty. \end{aligned}$$

And

$$\begin{aligned} \langle f\hat{g}, e^{in\theta} \rangle &= \langle f \cdot \sum_{k \in \mathbb{Z}} \hat{g}_k e^{ik\theta}, e^{in\theta} \rangle + \langle f \cdot \sum_{k \geq n} \hat{g}_k e^{ik\theta}, e^{in\theta} \rangle \\ &= \langle f, \sum_k e^{i(n-k)\theta} \hat{g}_k \rangle + \\ &= \sum_{k \in \mathbb{Z}} \hat{g}_k \hat{f}_{n-k} + \\ \Rightarrow \langle f\hat{g}, e^{in\theta} \rangle &= \sum_{k \in \mathbb{Z}} \hat{g}(k) \hat{f}(n-k). \end{aligned}$$

$\leq \|f\|_{L^2} \|\sum_{k \in \mathbb{Z}} \hat{g}_k\|$
 $\leq \|f\|_{L^2} \cdot \frac{1}{N^2} (\|\hat{g}\|_{L^2} + \|d\hat{g}\|_{L^2})$
 $\square$

Thm 20.8 : $\mathcal{F}(f * g) = \hat{f} \cdot \hat{g}$, 2π when all are defined.

Proof : $f * g = \int_0^{2\pi} f(\theta - \tau) g(\tau) d\tau$

$$\begin{aligned} \widehat{f * g}_n &= \frac{1}{2\pi} \int_0^{2\pi} \int_0^{2\pi} f(\theta - \tau) g(\tau) e^{-in(\theta + \tau - \tau)} d\tau d\theta \\ &= \frac{1}{2\pi} \int_0^{2\pi} \int_0^{2\pi} f(\theta - \tau) g(\tau) e^{-in\tau} e^{-in(\theta - \tau)} d\tau d\theta \end{aligned}$$

$$\begin{aligned} \theta' = \theta - \tau &= \frac{1}{2\pi} \int_0^{2\pi} \int_0^{2\pi} g(\tau) e^{-in\tau} f(\theta') e^{-in\theta'} d\theta' d\tau \quad \text{Fubini} \\ &= \int_0^{2\pi} \hat{g}(n) f(\theta') e^{-in\theta'} d\theta' = 2\pi \hat{f}(n) \hat{g}(n). \end{aligned}$$

Let $\delta \in C^0(S^1)^*$ be s.t. $\delta(f) = f(0)$. Written

$$\delta(f) := \frac{1}{2\pi} \int_0^{2\pi} \delta(x) f(x) dx.$$

Then

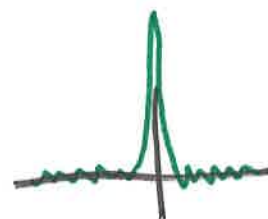
$$f(x) = \int_0^{2\pi} f * \delta(x) = \frac{1}{2\pi} \int_0^{2\pi} \delta(y) f(x-y) dy = f(x).$$

Intuitively,

$$\begin{aligned} \hat{F}(\delta) &= \frac{1}{2\pi} \int_0^{2\pi} \delta(x) e^{inx} dx \\ &= \frac{1}{2\pi} \forall n, \text{ and } \|\delta(x)\|_{L^2} = \infty \notin L^2 \end{aligned}$$



Def 20.9 : The Dirichlet Kernel for $N \in \mathbb{N}$

$$D_N(x) = \frac{1}{2\pi} \sum_{-N}^N e^{inx}$$


Thm 20.10 : 1) $\lim_{N \rightarrow \infty} f * D_N = f$ in L^2 $\forall f \in L^2$

(Approx Identity)

2) $f * D_N \in C^\infty$ and $\| \partial^m_{(f * D_N)} \|_{L^2} \leq N^m \|f\|_{L^2}$

Proof : 1) $\widehat{f * D_N} = \chi_{[-N, N]} \hat{f}$ so $\|f * D_N - f\|_{L^2} = \left\| \sum_{|n| > N} c_n e^{inx} \right\| \rightarrow 0$.

$$\begin{aligned} 2) \left\| \partial^m_{(f * D_N)} \right\|_{L^2}^2 &= \left| \sum |n|^{2m} \chi_{[-N, N]} \hat{f} \right|^2 \\ &\leq N^{2m} \left| \sum |n|^{2m} \hat{f} \right|^2 \leq N^{2m} \|f\|_{L^2}^2. \end{aligned}$$

and $\partial^m f \in L^2 \forall m \Rightarrow f$ is smooth.

Lecture 21: Convergence of Fourier Series (Fourier Series III)

21.1

Question 21.1: If $f \in L^2$ then $f_N = \sum_{|n| \leq N} \hat{f}(n) e^{in\theta} = f * D_N$ converges to f in L^2 .

Under what conditions does $f_N \rightarrow f$ pointwise? uniformly?

Ex 21.2: There exists an $f \in C^0(S')$ such that $f_N(0) \rightarrow \infty$.

(Fejér) $f(x) = \sum_{k=1}^{\infty} \sin((2^{k^3}+1)\frac{x}{2}) \cdot \frac{1}{k^2}$

Then $f(x)$ is a $\sup$ norm-limit of C^0 , so C^0 , but

$$\begin{aligned} \lim_{N \rightarrow \infty} f_N(0) &= \sum_{n=-\infty}^{\infty} a_n \\ &= \sum_{n=-\infty}^{\infty} \frac{1}{2\pi} \sum_{k=1}^{\infty} \frac{1}{k^2} \int_0^{\pi} \sin((2^{k^3}+1)\frac{\theta}{2}) \cos(n\theta) d\theta \\ &= \frac{1}{\pi} \sum_{n=1}^{\infty} \sum_{k=1}^{\infty} \frac{1}{k^2} \cdot \lambda_{n, 2^{k^3}+1} = \frac{1}{\pi} \sum_{k=1}^{\infty} \frac{\sigma_k}{k^2} \end{aligned}$$

Can show $\sigma_k = \sum_{n=1}^N \lambda_{n, 2^{k^3}+1}$ has $\sigma_k \geq \frac{1}{2} \log(N)$

$$\Rightarrow \text{For } N \geq 2^{k^3}-1, \quad f_N(0) \geq \frac{1}{k^2} \cdot \frac{1}{2} \log(2^{k^3}+1) \geq \frac{k}{L}.$$

21.i) Sobolev Spaces

Since $\widehat{-i\partial_{\theta} f} = n \hat{f}(n)$ we say

$$\partial_{\theta}^k f \in L^2 \iff n^k \hat{f} \in L^2$$

Define 21.3 the (continuous) resp (discrete) Sobolev norm of regularity k by

$$\|f\|_{L^{k,2}} = \left(\int_0^{2\pi} |f|^2 + |\partial f|^2 + |\partial^2 f|^2 + \dots + |\partial^k f|^2 d\theta \right)^{1/2}$$

$$\|\hat{f}\|_{\ell^{k,2}} = \left(\sum_{n=-\infty}^{\infty} (1 + |n|^2 + |n|^4 + \dots + |n|^{2k}) |\hat{f}(n)|^2 \right)^{1/2}.$$

Def 21.4

Define the Sobolev spaces

$$L^{k,2}(S'; \mathbb{C}), \quad \ell^{k,2}(\mathbb{Z}; \mathbb{C})$$

are the completions of C^∞ , finite seq in the above norms.

Thus $f \in L^{k,2}$ iff it has k -derivatives in L^2

* this needs "weak derivatives" to be made precise

Corollary 21.5 : The following diagram commutes

$$\begin{array}{ccc}
 L^2(S'; \mathbb{C}) & \xleftrightarrow{\mathcal{F}} & L^2(\mathbb{Z}; \mathbb{C}) \\
 \uparrow & & \uparrow \\
 L^{1,2}(S'; \mathbb{C}) & \xleftrightarrow{\mathcal{F}} & L^{1,2}(\mathbb{Z}; \mathbb{C}) \\
 \uparrow & & \uparrow \\
 L^{2,2}(S'; \mathbb{C}) & \xleftrightarrow{\mathcal{F}} & L^{2,2}(\mathbb{Z}; \mathbb{C}) \\
 \uparrow & & \uparrow \\
 \vdots & & \vdots
 \end{array}$$

and each horizontal arrow is an isometry in the $L^{k,2}$, $l^{k,2}$ norms.

Proof: inclusions commute w/ $\mathcal{F}$ so commuting is obvious.

As before

$$\begin{aligned}
 \int |f|^2 + |\partial f|^2 dx &= \sum_{n \in \mathbb{Z}} |\hat{f}(n)|^2 + |\partial \hat{f}(n)|^2 \\
 &= \sum_{n \in \mathbb{Z}} (1+n^2) |\hat{f}(n)|^2
 \end{aligned}$$

so $\mathcal{F}$ lands in $L^{1,2}$ precisely when $|\partial f| \in L^2$.

Caution 21.6 : $L^{k,2} \subseteq L^{k-1,2}$ is a (bounded) inclusion,

$$\|f\|_{L^{k-1,2}} \leq \|f\|_{L^{k,2}}$$

of a dense Banach space. So $L^{k,2}$ is closed in its own norm but dense in the $L^{k-1,2}$ -norm.

Thm 21.7: If $f \in L^{s,2}(S'; \mathbb{C})$, then $f_N \rightarrow f$ uniformly.

Proof: It suffices to show $\sup_{S'} |f| \leq C \|f\|_{L^{s,2}}$

Then, $f_N \rightarrow f$ in $L^{s,2}$ so $\sup_S |f_N - f| \leq C \|f_N - f\|_{L^{s,2}} \rightarrow 0$.
 ($\hat{f}_N \rightarrow \hat{f}$ in $L^{s,2}$ and isometry)

And $\leq \left| \sum \hat{f}(n) e^{in\theta} \right|$

$$\sup_{S'} |f| \leq \sum_{n \in \mathbb{Z}} |\hat{f}(n)|$$

$$\leq \sum_{n \in \mathbb{Z}} \frac{|\hat{f}(n)|}{\sqrt{1+n^2}} \sqrt{(n^2+1)}$$

$$\leq \left(\sum_{n \in \mathbb{Z}} |\hat{f}(n)|^2 (n^2+1) \right)^{1/2} \left(\sum_{n \in \mathbb{Z}} \frac{1}{n^2+1} \right)^{1/2}$$

$$\leq C \|\hat{f}\|_{L^{s,2}} = C \|f\|_{L^{s,2}}.$$

Corollary 21.8: If $f \in L^{k,2}(S'; \mathbb{C})$ then

$$\sup |f_N - f| + \sup |df_N - df| + \dots + \sup |d^{k-1} f_N - d^{k-1} f| \rightarrow 0.$$

Rem 21.9: $L^{s,2}$ makes sense for non-integer s , which gives a notion of fractional derivatives

$$\left(\frac{d}{ds} \right)^s f = (\mathcal{F}^{-1} \circ \ln^s \circ \mathcal{F}) f.$$

Actually the proof shows any $L^{s,2}$ for $s > \frac{1}{2}$ is okay.

21.iii) Fourier transform of L^p

21.4

Recall since S' is compact

$$L^2 \subseteq L^p \quad 1 \leq p \leq 2$$

$\parallel$

$$L^2 \subseteq L^{p'} \quad p' \geq 2.$$

Thm 21.10 (Hausdorff-Young)

Let $1 \leq p \leq 2$, and q be the Hölder conjugate.
 $\frac{1}{p} + \frac{1}{q} = 1.$

$$\text{Then } \|\mathcal{F}(f)\|_{L^q(\mathbb{Z})} \leq \|f\|_{L^p(S'; \mathbb{C})}$$

In particular,
 $\mathcal{F}: L^p(S'; \mathbb{C}) \rightarrow L^q(\mathbb{Z}; \mathbb{C})$
is bounded.

Proof: For $p=1$, $q=\infty$.

$$\|\hat{f}\|_{\infty} = \sup_n \left| \int \langle f, e^{inx} \rangle \right| \leq \|f\|_{L^1}.$$

$p=2$ is obvious. Interpolation or Young Convolution for general.

Ex 21.11: $\{1, 1, \dots\} = \mathcal{F}(\delta)$ in ℓ^∞ but $\delta \notin L^1$ so not surj.

$$\|D_N\|_{L^1} \approx \log N.$$

Thm 21.11: If $f \in L^p(S'; \mathbb{C})$ then for $1 < p < \infty$,

$$\|f_N - f\|_{L^p} \rightarrow 0.$$

Proof: "Convergence of Fourier Series in L^p Space" J. Miao, various Harmonic Analysis texts.

Lecture 22 | Fourier Series and PDEs I: Fredholm and Compact Operators (22.1)

Main Idea 22.1: $\mathcal{F}$ turns differentiation into multiplication

$$\begin{array}{ccc} L^{2,2}(S^1; \mathbb{C}) & \xleftrightarrow{\mathcal{F}} & \ell^{2,2}(\mathbb{Z}; \mathbb{C}) \\ \downarrow -\partial_x^2 & & \downarrow \ell^2 \\ L^2(S^1; \mathbb{C}) & \xleftrightarrow{\mathcal{F}} & \ell^2(\mathbb{Z}; \mathbb{C}) \end{array}$$

Commutates, eg.

$$L = \sum \alpha_j (-i\partial_x)^j f \longleftrightarrow (\sum \alpha_j n^j) \cdot \hat{f} \\ = (\alpha_N n^N + \alpha_{N-1} n^{N-1} + \dots)$$

$\Rightarrow$ Can solve $Lf = g$ by dividing Fourier transform

$$f = \mathcal{F}^{-1} \circ \frac{1}{|2|n^N + \alpha|2|n^{N-1} + \dots + \alpha_0} \cdot \mathcal{F}g := P_g$$

• As $|n| \rightarrow \infty$, only leading order matters

• Write $L = \underbrace{(|2|n^N + \alpha|2|n^{N-1} + \dots + \alpha_0)}_{\text{leading order} \approx \text{invertible}} + \underbrace{\dots}_{\text{lower order} \approx \text{Compact / f.d.}}$

$$\underbrace{\text{invertible} + \text{Compact / f.d.}}_{\text{Fredholm.}}$$

Question 22.2: Given $f \in X \subseteq L^2(S^1; \mathbb{C})$, when can we solve

$$Lu = f$$

and what space is u in?

22.1) : Compact Operators

(22-2)

Def 22.3 : Let H_1, H_2 be Hilbert spaces. An operator

$K: H_1 \rightarrow H_2$
is said to be compact if

$\{x_n\} \subseteq X$ bounded $\Rightarrow \{Kx_n\}$ has a convergent subsequence.

i.e. the image of the closed unit has compact closure.

Lemma 22.4 : If K has finite rank, then K is compact.

Proof : Let $E \subseteq H_2$ be $\text{Im}(K)$ and choose $E \cong \mathbb{R}^N$ some N .

If $\{x_n\}$ bounded by M ,

$$\|Kx_n\| \leq \|K\| \|x_n\| \leq \|K\| \|M\|.$$

So $\overline{\text{Im}(B_1)}$ is closed and bounded.

Lemma 22.5 : If $K_n \rightarrow K$ i.e. $\forall \varepsilon > 0 \exists N$ st $\|(K - K_n)x\| \leq \varepsilon \|x\|$
 $\forall n \geq N$
and K_n is finite rank, then K is compact.

Proof : Let $\{x_j\}$ be a bounded sequence

Let $\varepsilon > 0$, choose N so $\|K - K_N\| < \varepsilon/3$, and for that N , let J be large so $\|K_N x_j - K_N x_k\| < \varepsilon/3$ for $j, k \geq J$. Then for j, k

$$\begin{aligned} \|Kx_j - Kx_k\| &\leq \|Kx_j - K_N x_j + K_N x_j - K_N x_k + K_N x_k - Kx_k\| \\ &\leq \|Kx_j - K_N x_j\| + \|K_N x_j - K_N x_k\| + \|K_N x_k - Kx_k\| \\ &\leq \frac{\varepsilon}{3} + \frac{\varepsilon}{3} + \frac{\varepsilon}{3}. \end{aligned}$$

Rem 22.6 : This is iff (but not in Banach spaces!)

Lemma 22.7 : The unit ball $B \in H$ is compact iff H is finite dimensional. (22.3)

Proof : $\Leftarrow$ is Heine-Borel

$\Rightarrow$ if e_i is an infinite orthonormal basis $\|e_i - e_j\| = \sqrt{2}$ so no conv. subsequence.

Thm 22.8 (Rellich's Lemma) The inclusion $L^{k+1,2}(S'; \mathbb{C}) \subseteq L^{k,2}(S'; \mathbb{C})$

is compact.

Proof : Suffices to prove for $L^{1,2} \subseteq L^2$.

For each N let



$a_1, a_2, \dots \rightarrow x_1$ conv subsequence for $|e| \leq 1 \in \mathbb{R}^3$
 $b_1, b_2 \rightarrow x_2$ subsubseq for $|e| \leq 2$.

Claim diagonal is Cauchy. $\{y_n\}$ 1) For enough N , $\|y_n - y_m\| \leq \epsilon/2$ on $|e| \leq N$
 $= \sum_{i=1}^N |y_n - y_m|^2$

$$2) \sum_{|e| \geq N} |y_n - y_m|^2 \leq \frac{1}{N} \sum_{|e| \geq N} |e|^2 |y_n - y_m|^2 \leq \frac{2}{N}.$$

22.ii: Fredholm Operators

Def 22.9 : A bounded operator $L: H_1 \rightarrow H_2$ is said to have closed range if $L(H_1) \subseteq H_2$ is a closed subspace.

Lemma 22.10 (closed range estimate) $L: H_1 \rightarrow H_2$ has closed range iff $\exists C$ s.t.

$$\|x\|_{H_1} \leq C \|Lx\|_{H_2} \quad x \perp \text{Ker } L.$$

Proof : $\Rightarrow$ If $Lx_i \rightarrow y$ then $\|x_i - x_j\| \leq C \|Lx_i - Lx_j\| \rightarrow 0$
 so $x_i \rightarrow x$ and by continuity $Lx = y$.

$\Leftarrow$ Suppose $\|x_i\| = 1$ w/ $\|Lx_i\| \rightarrow 0$; then $\text{Im}(B) \subseteq \text{Ker}(L)^\perp$ is closed, and contains 0, so $\exists x \in B$ w/ $Lx = 0 \rightarrow \Leftarrow \square$

Def 22.11 : An operator $L: H_1 \rightarrow H_2$ is said to be Fredholm if

- L has f.d. kernel
- L has closed range
- L has f.d. cokernel.

Rem 22.12 : Closed range is key! $H_2 / \text{Im}(L) \cong H_2 / \overline{\text{Im}(L)} \approx \text{Im}(L)^\perp$

eg $L^2 / L^{1,2}$ is a bit hard to parse

Thm 22.13 : An operator $L: H_1 \rightarrow H_2$ is Fredholm iff $\exists P: H_2 \rightarrow H_1$ st
 $LP = \text{Id} + K_1$ $PL = \text{Id} + K_2$
 For $K_i: H_i \rightarrow H_i$ compact.

Proof : $\Rightarrow$ Suppose L Fredholm. Let P be inverse of $\text{Ker}(L)^\perp \xrightarrow{P} \text{Im}(L)$ (bounded by open mapping thm)

Then $PL = \text{Id} - \pi_{\text{Ker } L}$ which is compact b/c f.d. finite rank
 $\forall LP = \text{Id} - \pi_{\text{Im } L}$ "

$\Leftarrow$ $0 = PL(x_i) = x_i - Kx_i$ for $\|x_i\|=1$ in $\text{Ker } L$.
 $\|x_i\| \leq \|Kx_i\|$ but converges of x_i so x_i converges $\rightarrow B \cap \text{Ker } L$ is compact, so f.d.

For closed range

$$\|x\| \leq \|PLx + Kx\| \leq C\|Lx\| + \|Kx\|$$

Then $LP(y_i) = y_i + Ky_i$
 so $y_i = Ky_i$ on $\text{Im } L^\perp$ same.

Now suppose $x \perp \text{Ker } L$ If $Lx_i \rightarrow y$ then

$$\|x_i - x_j\| \leq C(\|L(x_i) - Lx_j\| + \|Kx_i - Kx_j\|) \text{ on subseq.}$$

Corollary ^{22.14} : Fredholm implies $\exists C, K$ compact so $\rightarrow 0$

$$\|x\|_{H_1} \leq C(\|Lx\|_{H_2} + \|Kx\|_{H_1})$$

Lecture 23 Fourier Series and PDEs II: Elliptic Equations

23.1

Recall the original goal was to study solutions of

$$Lu = f$$

For differential operators L .

Question 23.1 : For what f can this be solved?

What is $\text{Ker}(L) \subseteq L^{k,2}$?

Question 23.2 : For $f \in L^{k,2}$, what space does u lie in?

Recall $\Delta = -\partial_\theta^2 = (-i\partial_\theta)^2$

We will consider four examples

i) $L = \Delta + 1$

ii) $L = \Delta$

iii) $L = \sum_{j=0}^N \alpha_j (-i\partial_\theta)^j \quad \alpha_j \in \mathbb{C}$

iv) $L = \Delta + V(\theta)$ (time-ind Schrödinger)

Def 23.3 : An operator $L = \sum_{j=0}^N \alpha_j(x) (-i\partial_\theta)^j$ is called a differential operator for $\alpha_j : S^1 \rightarrow \mathbb{C} \in C^\infty$. It is said to be elliptic if $|\alpha_N(x)| > 0$.

Observe that an N^{th} order operator is a bounded map

$$L : L^{k+N,2}(S^1; \mathbb{C}) \rightarrow L^{k,2}(S^1; \mathbb{C})$$

since $\|Lu\|_{L^2} \leq \left\| \sum_{j=0}^N \alpha_j(x) (-i\partial_\theta)^j u \right\|_{L^2} \leq \sum_{j=0}^N \|\partial_\theta^j u\|_2 \leq \|u\|_{L^{N,2}}$

same for $L^{k,2}$.

23.i) $L = \Delta + 1$

$$L^{2,2}(S^1; \mathbb{C}) \xrightarrow{\mathcal{F}} L^{2,2}(S^1; \mathbb{C})$$

$$\begin{array}{ccc} \Delta+1 \downarrow & & \downarrow |e|^2+1 \\ L^2(S^1; \mathbb{C}) & \xrightarrow{\mathcal{F}} & L^2(S^1; \mathbb{C}) \end{array}$$

Prop 23.4 : $\Delta + 1 : L^{2,2}(S^1; \mathbb{C}) \rightarrow L^2(S^1; \mathbb{C})$ is an isomorphism

Proof : Suffices to show $|e|^2 + 1 : L^{2,2} \rightarrow L^2$ is, but this obviously has inverse $\frac{1}{|e|^2 + 1}$.

To be pedantic,

23.2

$$\begin{aligned} \|\frac{1}{(1+|x|^2)^{k/2}} a\|_{L^{k,2}} &= \sum_{\ell \in \mathbb{Z}} (1+|\ell|^2+|\ell|^4) \frac{|a_\ell|^2}{(1+|\ell|^2)^2} \\ &= \sum_{\ell \in \mathbb{Z}} \frac{1+|\ell|^2+|\ell|^4}{1+2|\ell|^2+|\ell|^4} |a_\ell|^2 \leq C \|a\|_{L^2}^2. \end{aligned}$$

Corollary 23.5: 1) $\Delta+1 : L^{k+2,2}(S'; \mathbb{C}) \rightarrow L^{k,2}(S'; \mathbb{C})$ is an isomorphism $\forall k \in \mathbb{N}$.

2) $\Delta+1$ (a fortiori) is Fredholm $\forall k$,

3) $\Delta+1$ satisfies

$$\|u\|_{L^{k,2}} \leq C_k \|(\Delta+1)u\|_{L^{k,2}} \text{ for } C_k \text{ constants.}$$

4) $Lu = f$ has a solution $u, \forall f \in L^{k,2}$ and $u \in L^{k+2,2}$.

Proof: Only 1) needs proof

$$\begin{aligned} (1+|\ell|^2 + \dots + |\ell|^{2k} + |\ell|^{2k+2} + |\ell|^{2k+4}) &\sim 1+|\ell|^{2k+4} \\ &\sim (1+|\ell|^{2k})(1+|\ell|^2)^2 \end{aligned}$$

Remark 23.6: Actually the same result holds for $s \in \mathbb{R}$.

23.ii) $L = \Delta$

Def 23.7: The (formal) adjoint of an operator L is that L^* such that

$$\langle Lu, v \rangle_{L^2} = \langle u, L^* v \rangle_{L^2}$$

$\forall u, v \in \mathcal{C}^\infty$. L is said to be formally self-adjoint if $L = L^*$.

Ex 23.8:
$$\begin{aligned} \langle Lu, v \rangle &= \int \sum (\alpha_j(x) \partial_j) \bar{\partial}_j u, v \rangle \\ &= \int \langle u, (-i \partial_j)^{n_j} \bar{\alpha}_j(x) v \rangle \\ &= \bar{\alpha}_j(x) (-i \partial_j)^N + \bar{\alpha}_{N-1} (-i \partial_j)^{N-1} + (\partial \alpha_N) (-i \partial_j)^{N-1} \end{aligned}$$

Ex 23.9 Δ is formally self adjoint.

[23.3]

$$\int \langle -\partial_{\bar{z}}^2 u, v \rangle = \int \langle -\partial_{\bar{z}} u, -\partial_{\bar{z}} v \rangle = \int \langle u, -\partial_{\bar{z}}^2 v \rangle.$$

Thm 23.10 : The operator

$$\Delta : L^{k+2,2}(S'; \mathbb{C}) \rightarrow L^{k,2}(S'; \mathbb{C})$$

is Fredholm. It has 1-dimensional kernel and cokernel, given by the constants.

Proof 1: On Fourier space $\Delta = |\ell|^2$. Set $P = \begin{pmatrix} \frac{1}{|\ell|^2} & \ell \neq 0 \\ 0 & \ell = 0 \end{pmatrix}$.

$$\text{Then } P\Delta = \text{Id} - \pi_0$$

$$\Delta P = \text{Id} - \pi_0$$

and π_0 has rank 1, so compact.

Therefore, Δ is Fredholm. Clearly constants are both the kernel and cokernel.

~~Proof 2 (of cokernel)~~ Corollary 23.11 (Elliptic estimates)

$$\|u\|_{L^{2,2}} \leq \|\Delta u\|_{L^2} + \|\pi_0 u\|_{L^2} \leq \|\Delta u\|_{L^2} + \|u\|_{L^2}$$

$$\|u\|_{L^{k+2,2}} \leq \|\Delta u\|_{L^{k,2}} + \|u\|_{L^{k,2}}$$

Corollary 23.12 (Elliptic regularity). If $\Delta u = f$ w/ $f \in C^\infty$ (in particular 0) then $u \in L^2 \Rightarrow u \in L^{2,2} \Rightarrow u \in L^{4,2} \dots \Rightarrow u \in C^\infty$.

Proof : $\|u\|_{L^{k+2,2}} \leq \|\Delta u\|_{L^{k,2}} + \|u\|_{L^{k,2}}$ same for f .

Remark 23.13 : "u is 2 degrees more regular than itself"
This is called "elliptic bootstrapping".

Proof 2 (of cokernel) : If $v \in \text{Coker}$, then

$$0 = \langle \Delta u, v \rangle = \langle u, \Delta v \rangle \text{ for all } u \Rightarrow \Delta v = 0 \text{ and so } v \in L^{2,2}.$$

Therefore $v \in \ker(\Delta)$, so constant.

iii) General Const Coefficient

(23.4)

Thm 23.14: $L = \sum_{j=1}^N a_j (-i\partial)^j$ is Fredholm, it satisfies

$$\|u\|_{L^{k+2,2}} \leq C_k \|Lu\|_{L^{k,2}} + \|u\|_{L^{k,2}}$$

and elliptic regularity. It has ~~ker~~ $\dim \ker = \dim \operatorname{Coker} = \left\{ \begin{array}{l} \# \text{ integer roots} \\ \text{of } \sum a_j \lambda^j = 0 \end{array} \right\}$

Proof: $\mathcal{F}^{-1} L \mathcal{F} = a_0 + a_1 i + a_2 i^2 + \dots + a_N i^N$.

Rem 23.15: Agrees w/ ODE theory that N^{th} order has $\leq N$ solutions.

iv) The time-independent Schrödinger Equation

Say $V(x) \in C^\infty$ w/ $V \geq 0$.

$$\Delta + V \leftrightarrow |\xi|^2 + \hat{V}^* \quad \leftarrow \text{ew.}$$

Thm 23.16: $\Delta + V$ is Fredholm and formally self adjoint.
 $\exists C_k$ such that

$$\|u\|_{L^{k+2,2}} \leq C_k \|(\Delta + V)u\|_{L^{k,2}} + \|u\|_{L^{k,2}}$$

hold.

Proof: $\|u\|_{L^{k+2,2}} \leq \|\Delta u\|_{L^{k+2,2}} + \|u\|_{L^{k,2}}$
 $\leq \|(\Delta + V - V)u\|_{L^{k,2}} + \|u\|_{L^{k,2}}$
 $\leq \|(\Delta + V)u\|_{L^{k,2}} + \|u\|_{L^{k,2}} \quad C_k \sim \|V\|_{C^k}$

Thm 23.17: In fact, $\operatorname{Ker}(\Delta + V) = \{0\}$.

~~It's not, it's not~~. Can show $\|u\|_{L^2} \leq \|\Delta + V\|_{L^2}$ then $\|u\|_{L^2} \leq C \|\Delta + V\|_{L^2}$
 If not $\exists u_n$ st $\|u_n\|_{L^2} = 1$, but $\|\Delta + V\|_{L^2} \leq \frac{1}{n}$.

Then 1) ~~It's not, it's not~~ $\langle (\Delta + V)u, u \rangle \leq \frac{1}{\sqrt{n}} \|u_n\|_{L^2}^2 + \sqrt{n} \|\Delta + V\|_{L^2} \|u_n\|_{L^2}^2 \rightarrow 0$
 $\| \nabla u_n \|_{L^2}^2 + \langle u, Vu \rangle_{L^2} \rightarrow 0$.

2) $u_n \in L^2$, so $\|u_n\|_{L^{2,2}} \leq \frac{1}{n} + \|u_n\|_{L^2} < \infty$.
 $\leq C_{u_2}$ bounded

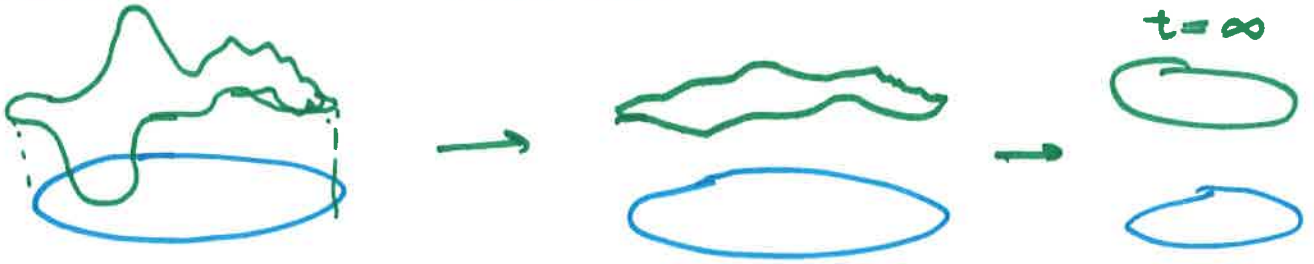
$\Rightarrow u_{n_k} \in L^{1,2}$ convergent, but $\|Vu_{n_k}\| \rightarrow 0$ so $u_\infty = \text{const}$,
 $\rightarrow u_\infty$ since $\|u_n\|=1$,

but $\langle u, Vu \rangle = \frac{1}{8\pi} \int v > 0. \rightarrow \leftarrow$
 $\nRightarrow u_\infty = \frac{1}{2\pi}.$
 $\square.$

Lecture 24 Fourier Series and PDEs III: the heat equation and 24.1 heat kernel.

Question 24.1

Suppose that $f(x) \in C^\infty(\mathbb{S}^1; \mathbb{R})$ is a smooth heat distribution at time $t=0$. What is the distribution of heat $u(x,t)$ for all t ?



Def 24.2: the initial value problem for the heat equation is to solve

$$\begin{cases} (\partial_t + \Delta_x) u(x,t) = 0 & = g(x,t) \text{ heat source} \\ u(x,0) = f(x). \end{cases}$$

for $t \geq 0$.

Ex 24.3: Suppose that $f(x) = \cos(nx)$. Then

$$\Delta_x f = -\partial_x^2 f = n^2 f \quad \text{so}$$

$$(\partial_t + \Delta_x) e^{-n^2 t} \cos(nx) = (-n^2 + n^2) e^{-n^2 t} \cos(nx) = 0.$$

is a solution.

Thm 24.4 (general solution) The heat evolution preserves Fourier modes and if

$$f(x) = \sum_{k \in \mathbb{Z}} a_k \cos(kx) + b_k \sin(kx)$$

then

$$u(x,t) = \sum_{k \in \mathbb{Z}} e^{-k^2 t} (a_k \cos(kx) + b_k \sin(kx))$$

solves

$$\begin{cases} (\partial_t + \Delta_x) u(x,t) = 0 \\ u(x,0) = f(x). \end{cases}$$

Lemma 24.5: If $f(x) \in L^2(S; \mathbb{R})$ then $u(x, t) \in C^\infty(S; \mathbb{R})$ [24.7]
for each $t > 0$.

Proof: It suffices to show that $\sum |u_k|^2 |k|^{2s} < \infty$ for all s .

But $u_k = \frac{e^{-k^2 t}}{(a_k + ib_k)^{1/2}}$ so $\sum (|a_k|^2 + |b_k|^2) |k|^{2s} e^{-k^2 t} \leq C \sum |a_k|^2 + |b_k|^2 \leq C \|f\|_{L^2}^2$
because $|k|^{2s} e^{-k^2 t} < 1$ for k large.

Proof (of thm 24.4)

By the above $\|u\|_{L^2, 2} \neq \infty$, so

$$\Delta_x \left(\sum_{|k|=1}^N e^{-k^2 t} (a_k \cos(kx) + b_k \sin(kx)) \right) \xrightarrow{u_N} \Delta u.$$

For each $t \in [t_0, t_1]$ the bound is uniform,

$$\partial_t \left(\sum_{|k|=1}^N e^{-k^2 t} (a_k \cos(kx) + b_k \sin(kx)) \right) \rightarrow \partial_t u \text{ uniformly}$$

(use large s , so bringing down k^2 doesn't matter).

Then by taking limits

$$(\partial_t + \Delta_x) u = (\partial_t + \Delta_x) u_N = 0 \quad \text{by previous example. } \square$$

obviously $u(x, 0) = f$

Corollary 24.6: $u(x, t) \in C^\infty((t_0, \infty) \times S; \mathbb{R})$ for any t_0 .

Proof: Lemma 24.5 shows x derivatives exist.

For t derivatives, $k^{2s} u_N \rightarrow u$ and

$$k^{2s} u_N' \rightarrow v \text{ uniformly}$$

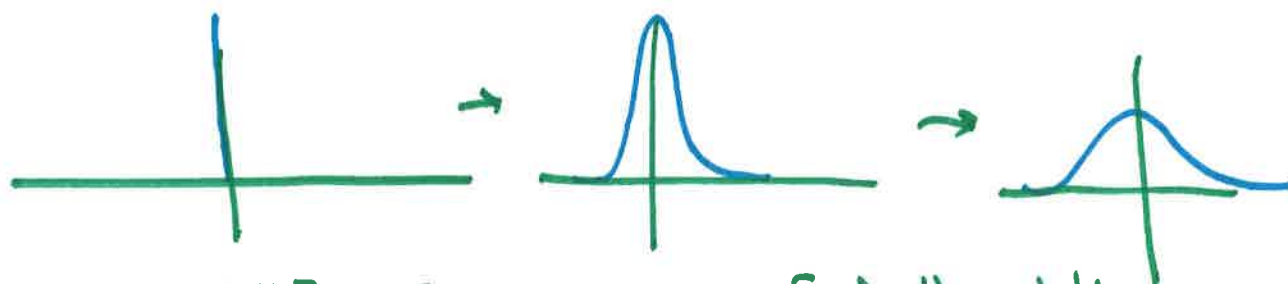
$$\text{for any } s, \text{ thus } \partial_t^s u = \partial_t^s u_N = k^{2s} u_N.$$

Mixed partials same. □

24.ii) The idea of a fundamental solution

24.3

Consider a pinprick of heat



Observation 24.7: Suppose we can find the solution of $\begin{cases} (\partial_t + \Delta_x) u(x,t) = 0 \\ u(x,0) = \delta_x. \end{cases}$

Then $f(x) \approx \sum \delta(x - \frac{j}{n}) f(\frac{j}{n}) \rightarrow \int_0^{2\pi} \delta(x-y) f(y) dy.$

the heat equation is Linear so

$$u_t(x,t) = \sum_{j=1}^N K(x - \frac{j}{n}, t) \cdot f(\frac{j}{n}) \rightarrow \int_0^{2\pi} K(x-y, t) f(y) dy.$$

Def 24.8: the solution $K(x-y, t) : S^1 \times S^1 \times \mathbb{R}^{\geq 0} \rightarrow \mathbb{R}$ is called the heat kernel or fundamental solution.

Thm 24.9: There exists a heat kernel $K(x,t)$ such that $(\partial_t + \Delta_x) K(x,t) = 0 \quad \forall t > 0$

and as $t \rightarrow 0$,

$$\int_0^{2\pi} K(x,t) dx \rightarrow 1 \quad \text{as } t \rightarrow 0$$

$$\int_{\delta}^{2\pi-\delta} K(x,t) dx \rightarrow 1 \quad \forall \delta > 0, \exists t \text{ such that}$$

$$\int_0^{2\pi} K(x,t) f(y) dy \rightarrow f(\frac{x}{n}) \text{ uniformly as } t \rightarrow 0.$$

Proof: Recall the Dirichlet Kernel

$$D_N(x) = \sum_{j=-N}^N e^{in\theta}.$$

then is $\hat{D}_N = \{ \delta_{k,0} \}_{k \in \mathbb{Z}}$.

Thus set

$$K(x,t) = \sum_{k \in \mathbb{Z}} e^{-k^2 t} \cos(kx).$$

By the same logic as before, $K(x,t) \in C^\infty((0,\infty) \times S^1; \mathbb{R})$ and $(\partial_t + \Delta_x) K(x,t) = 0$.

Lemma 24.10 (Poisson Summation)

proved next week.

$$\sum_{k \in \mathbb{Z}} f(x+k) = \sum_{n \in \mathbb{Z}} e^{in\theta} \hat{f}(n)$$

In particular, since $\int_{\mathbb{R}} \mathcal{F}(e^{-ax^2}) = e^{-x^2/a}$, the following results

$$K(x,t) = \sum_{k \in \mathbb{Z}} e^{-k^2 t} \cos(kx) = \sum_{k \in \mathbb{Z}} \frac{1}{\sqrt{4\pi t}} e^{-\frac{(x-k)^2}{4t}}$$

~~bullet~~ bullet points 1), 2) obvious from Fourier, real space side resp.
For 3) know D_N satisfies this, and $K_N \rightarrow D_N$ uniformly. $\square$

Rem 24.11: Kernels are used as models for the linear parabolic theory underlying "flow" equations e.g. Mean Curvature flow, Ricci flow, Yang-Mills flow.

It's not a coincidence that $\int K(x,t) f(y) dy \xrightarrow{t \rightarrow \infty}$

$$\lim_{t \rightarrow \infty} \text{Tr}(f \mapsto \int K(x,y,t) f(y)) = 1 = \dim \text{Ker } \Delta$$

This is used in Getzler's famous Heat kernel proof of the Atiyah-Singer Index theorem.

Lecture 25 Fourier Transforms I: Schwartz functions + distributions.

Question 25.1: What is the analogue of Fourier series on $\mathbb{R}$?

Recall functions on $\mathbb{R}$ are more subtle b/c we must control the behavior as $|x| \rightarrow \infty$.

On $[-\pi, \pi]$, write $f(x) = \sum_k a_k e^{ikx}$ periodic functions
 $[-\pi, \pi]$ $f(x) = \sum_k a_k e^{i \frac{k}{T} x}$ for $\frac{k}{T} = 0, \frac{1}{T}, \frac{2}{T}, \dots$

Note the discretization of the series gets



finer as $T \rightarrow \infty$. Write $\xi_k = \frac{k}{T}$ then

$$g(\xi_k) = \frac{1}{2\pi} \int_{-\pi/2}^{\pi/2} f(x) e^{i \xi_k x} dx$$

$$f(x) = \frac{1}{\pi T} \sum g(\xi_k) e^{i x \xi_k} = \sum g(\xi_k) e^{i x (\xi_{k+1} - \xi_k)} \\ \rightarrow \frac{1}{2\pi} \int_{-\pi/2}^{\pi/2} e^{i x \xi} g(\xi) d\xi$$

In limit, $g = \hat{f}$ becomes a function of continuous variable ξ .

Rem 25.2: Fourier series has nice properties

1) Countable basis

2) $-i\partial_x \leftrightarrow k$

3) $\|f(\omega)\|_{L^2} = \|\hat{f}\|_{L^2}$

$\Leftrightarrow$

Transform or Spectral

X

✓

✓

✓

X

X

Observation 25.3: It is no longer clear $\hat{f}$ is integrable, or $\mathcal{F}, \mathcal{F}^{-1}$ make sense etc.

Def 25.34: Given f in some function space X , define the Fourier transform

by $\hat{f} = \mathcal{F}(f) = \frac{1}{2\pi} \int_{\mathbb{R}} f(x) e^{i \xi x} dx$

and the inverse transform by

$$g = \mathcal{F}^{-1}(\hat{f}) = \frac{1}{2\pi} \int_{\mathbb{R}} \hat{f}(\xi) e^{-i \xi x} d\xi$$

25.i: Schwartz Functions

Def 25.5: A function is said to be Schwartz if for all $\alpha \in \mathbb{N}$, $N \in \mathbb{N}$

$$|x^N \partial_x^\alpha f| \leq C_{N,\alpha}$$

i.e. functions whose derivatives all decay faster than polynomially.

Ex 25.6: e^{-x^2} is Schwartz.

Prop 25.7: $\mathcal{S}(\mathbb{R})$ the Schwartz functions satisfies

- 1) is a vector space and algebra
- 2) $\partial_x, \int_{-\infty}^x : \mathcal{S}(\mathbb{R}) \rightarrow \mathcal{S}(\mathbb{R})$, and $e^{isx} : \mathcal{S}(\mathbb{R}) \rightarrow \mathcal{S}(\mathbb{R}) \quad \forall s \in \mathbb{R}$
- 3) ~~obviously~~ $\mathcal{S}(\mathbb{R}) \subset L^\infty(\mathbb{R}) \cap L^2(\mathbb{R}) \cap L^1(\mathbb{R})$.

Proof: these are all obvious.

Warning 25.8: $\mathcal{S}(\mathbb{R})$ is not a Banach or Hilbert space. The topology is a bit more complicated, and stronger.

$$\varphi_n \rightarrow \varphi \quad \text{iff} \quad |x^N \partial_x^\alpha (\varphi_n - \varphi)| \rightarrow 0 \quad \forall N, \alpha$$

It is what called a ^(Fischer) Fréchet space w a countable filtration of norms.

Lemma

Def 25.9: For each $\varphi \in \mathcal{S}(\mathbb{R})$, $\hat{\varphi}$ is defined and $\hat{\varphi} \in L^\infty(\mathbb{R})$.

$$\hat{\varphi}(\xi) = \frac{1}{2\pi} \int_{\mathbb{R}} f(x) e^{isx} dx \leq \frac{1}{2\pi} \int_{\mathbb{R}} |f| dx = \|f\|_1 < \infty.$$

So take sup.

25.ii) Properties of the Fourier transform

Prop 25.10: The Fourier transform on $\mathcal{S}(\mathbb{R})$ obeys

- 1) $\widehat{f(x-y)} = e^{-is y} \hat{f}(\xi)$, $\widehat{e^{isx} f(x)} = \hat{f}(\xi + \zeta)$
- 2) $\widehat{\partial_x^k f} = (-i\xi)^k \hat{f}$

$$3) \widehat{(ix)^k f(x)} = \partial_\xi^k \hat{f}(\xi).$$

$$4) \widehat{f * g} = \hat{f} \cdot \hat{g} \quad (\text{inverse hard to show directly.})$$

Proof:

$$1) \widehat{f(x+y)} = \int_{\mathbb{R}^n} f(x+y) e^{i\xi x} dx = \int_{\mathbb{R}^n} f(x) e^{i(x-y)\xi} dx = e^{-i y \xi} \hat{f}(\xi).$$

$$\begin{aligned} 2) \widehat{(\partial_x^k f)} &= \int_{\mathbb{R}^n} (\partial_x^k f) e^{i\xi x} dx \\ &= (-1)^k \int_{\mathbb{R}^n} f \partial_x^k (e^{i\xi x}) dx \quad \text{Int by parts.} \\ &= (-i\xi)^k \int_{\mathbb{R}^n} f e^{i\xi x} dx \end{aligned}$$

$$3) \frac{\partial}{\partial \xi} \hat{f} = \lim_{h \rightarrow 0} \int_{\mathbb{R}} f(x) \frac{e^{+i(\xi+h)x} - e^{i\xi x}}{h} dx$$

$$= \int_{\mathbb{R}} f(x) (ix) e^{i\xi x} dx$$

$$= \widehat{(ix)f(x)} \quad \text{and induct.}$$

↓ dominated convergence

4) same as series.

Corollary 25.11 : $\mathcal{F}: \mathcal{D}(\mathbb{R}) \rightarrow \mathcal{D}(\mathbb{R})$
 $\mathcal{F}^{-1}: \mathcal{D}(\mathbb{R}) \rightarrow \mathcal{D}(\mathbb{R})$

Proof: As before, if $\varphi \in \mathcal{D}(\mathbb{R})$, $|\hat{\varphi}| < \infty$.

and $|\xi^N \partial_\xi^k \hat{\varphi}| = |\widehat{x^k \partial_x^N \varphi}| < \infty$ b/c $x^k \partial_x^N \varphi \in \mathcal{D}(\mathbb{R})$

The same properties hold w/ $\mathcal{F}^{-1}$ but w/ $(-1)^{N+k}$. $\square$

Thm 25.12 : $\mathcal{F} : \mathcal{S}(\mathbb{R}) \rightarrow \mathcal{S}(\mathbb{R})$ is an isomorphism with inverse $\mathcal{F}^{-1}$.

Proof :
$$\mathcal{F}^{-1} \circ \mathcal{F} (x^N \partial_x^k \varphi) = \mathcal{F}^{-1} (\partial_y^N \zeta^k \mathcal{F}(\varphi))$$
$$= x^N \partial_x^k \mathcal{F}^{-1} \mathcal{F}(\varphi).$$

Thus $T = \mathcal{F}^{-1} \circ \mathcal{F} : \mathcal{S} \rightarrow \mathcal{S}$ commutes w/ x, ∂_x .

Lemma 25.13 : If $[T, x], [T, \partial_x] = 0$ on $\mathcal{S}$, then $T = c \text{Id.}$ for $c \in \mathbb{R}$.

Proof: Step 1 : If $\varphi(y) = 0$ then $T\varphi(y) = 0$.

$\varphi = (x-y)\varphi_1$ w/ $\varphi_1 \in \mathcal{S}$ by Taylor's theorem.

Thus
$$T\varphi(y) = T(x-y\varphi_1)(y)$$
$$= [(x-y)T\varphi_1](y) = 0.$$

Step 2 : Fix $y \in \mathbb{R}$, g st $g(y) = 1$ st $c(y) = (Tg)(y)$.

Then $Tf(y) = c(y)f(y)$. To see this,

$\varphi_2 = f(y) - f(y)g(x)$ so $\varphi_2(y) = 0$,

so $0 = T\varphi_2 = Tf(y) - f(y)(Tg)(y)$
$$= Tf(y) - f(y)c(y) \text{ so } Tf(y) = c(y)f(y) \quad \forall y.$$

Step 3 : Take $f \neq 0$ eg e^{-x^2} , then $c(y) = \frac{Tf}{f} \in \mathcal{S}(\mathbb{R})$.

Step 4 : $c(y)(\partial_x f)_y = T(\partial_x f)_y = \partial_x T f(y)$
$$= \partial_x [c \cdot f](y).$$
$$= (\partial_x c)_y \cdot f(y) + c(y) \partial_x f(y)$$

thus $(\partial_x c)(y) = 0 \quad \forall y \Rightarrow c$ is const.

Step 5 : By normalizing conventions, can take $c=1$.

Lecture 26 Fourier Transforms II: Distributions and Plancherel.

[26.1]

Recall that if $E \subseteq \mathbb{R}^n$ is compact

$$L^\infty(E) \subseteq \dots \subseteq L^4(E) \subseteq L^3(E) \subseteq L^2(E) \subseteq L^1(E)$$

$\Downarrow$ * dual

$$\dots \supseteq L^{4/3}(E) \supseteq L^{3/2}(E) \supseteq L^2(E) \supseteq L^1(E)$$

Thus in general $X \subseteq Y \Rightarrow Y^* \subseteq X^*$ (all functionals bounded on Y are automatically bounded on X , but maybe not).

Def 26.1: The space of tempered distributions, denoted $\mathcal{S}'(\mathbb{R}) = \mathcal{S}(\mathbb{R})^*$ is $\mathcal{S}'(\mathbb{R}) = \{ \phi: \mathcal{S}(\mathbb{R}) \rightarrow \mathbb{R} \mid \phi \text{ is bounded in the sense } \exists C, N, K \text{ s.t. } |\phi(\varphi)| \leq C \sum_{\substack{K, n \\ \leq K, N}} \sup_{\mathbb{R}} |x^n \partial_x^K \varphi| \}$

Ex 26.2: $\alpha = |x|^N$. Then

$\alpha \cdot \varphi = \int_{\mathbb{R}} |x|^N \cdot \varphi dx = \langle \alpha, \varphi \rangle$ is a tempered distribution as

$$\int_{\mathbb{R}} |x|^N \varphi = \int_{\mathbb{R}} |x|^N \varphi \leq \sup_{\mathbb{R}} |x|^N |\varphi| \text{ so } N \leq N.$$

Ex 26.3: $|x|^n \partial_x^K = \frac{1}{i} \partial_x^{K-1} |x|^n$

$$\varphi \mapsto \int_{\mathbb{R}} |x|^n \partial_x^K \varphi dx \text{ for same reason}$$

$\Rightarrow$ functions w/ growth slower than some polynomial.

Ex 26.4: Let $\delta \in \mathcal{S}'(\mathbb{R})$ be the distribution given by

$$\langle \delta, \varphi \rangle = \int_{\mathbb{R}} \delta(x) \varphi(x) dx = \varphi(0).$$

Def 26.5: δ is called the Dirac delta "function" or distribution.

26.7

26.ii) Operators on distributions

Suppose $A: \mathcal{D}(\mathbb{R}) \rightarrow \mathcal{D}(\mathbb{R})$ is a linear (continuous) operator.

Def 26.6: The formal adjoint is the operator on distributions

$A^*: \mathcal{D}(\mathbb{R}) \rightarrow \mathcal{D}(\mathbb{R})$ given by

$$\langle A^* u, \varphi \rangle := \langle u, A\varphi \rangle$$

Lemma 26.7: $\forall u \in \mathcal{D}'(\mathbb{R}), \exists \varphi_n \in \mathcal{D}(\mathbb{R})$ s.t.

$$\int_{\mathbb{R}} \langle \varphi_n, \varphi \rangle dx = \langle \varphi_n, \varphi \rangle \longrightarrow u(\varphi) \text{ for all } \varphi.$$

Proof: Take $\varphi_n(x) = \int K_{1/n}(x-y) u(y) dy$
 $= \langle K_{1/n}(x-y), u \rangle$
and apply approximate identity stuff.

Def 26.8: A^* is said to extend B if

$$\langle A^* u, \varphi \rangle = \langle u, A\varphi \rangle = \langle Bu, \varphi \rangle$$

for all $u, \varphi \in \mathcal{D}'(\mathbb{R})$

Ex 26.9: The distributional derivative of u is

$$\langle \partial_x u, \varphi \rangle := \langle u, -\partial_x \varphi \rangle.$$

Obviously, if $u \in \mathcal{D}(\mathbb{R})$ this holds by integration by parts, so the extension is valid.

Ex 26.10: $H(x) = \begin{cases} 1 & x \geq 0 \\ 0 & x < 0 \end{cases}$. $-\delta = \partial_x H$. Since $\forall \varphi$,

$$\langle H, \partial_x \varphi \rangle = \int_0^\infty \partial_x \varphi = \varphi(\infty) - \varphi(0) = -\varphi(0)$$

26.iii) The Plancherel Theorem

(26.iii)

Let $A: \mathcal{L}(\mathbb{R}) \rightarrow \mathcal{L}(\mathbb{R})$ be the Fourier transform. Thus to extend to $\mathcal{L}'(\mathbb{R})$ we need a $B: \mathcal{L}(\mathbb{R}) \rightarrow \mathcal{L}(\mathbb{R})$ satisfying

$$\int_{\mathbb{R}} (B\psi) \varphi dx = \langle B\psi, \varphi \rangle = \langle \psi, A\varphi \rangle = \int_{\mathbb{R}} \psi(x) \hat{\varphi}(x) dx$$

Thm 26.11 (Plancherel): $B = \mathcal{F}$. i.e., the Fourier transform is self-adjoint on S . i.e.

$$\int_{\mathbb{R}} \hat{\psi} \varphi dx = \int_{\mathbb{R}} \psi \hat{\varphi} dx$$

Proof :

$$\begin{aligned} \int_{\mathbb{R}} \hat{\psi} \varphi dx &= \int_{\mathbb{R}} \int_{\mathbb{R}} \psi(\xi) e^{i\xi x} d\xi \varphi(x) dx \\ &= \int_{\mathbb{R}} \int_{\mathbb{R}} \psi(\xi) e^{i\xi x} \varphi(x) dx d\xi \quad (\text{Fubini}) \\ &= \int_{\mathbb{R}} \psi(\xi) \hat{\varphi}(\xi) d\xi. \end{aligned}$$

Corollary 26.12: $\mathcal{F}$ extends to a map $\mathcal{F}: \mathcal{L}'(\mathbb{R}^n) \rightarrow \mathcal{L}'(\mathbb{R}^n)$ as does $\mathcal{F}^{-1}$, and they remain inverses.

Proof: Follows directly, and extension of id is obviously id.

Corollary 26.13: $\mathcal{F}$ extends to an isomorphism on L^2 , in fact an isometry.

So

$$\begin{array}{ccccc} \mathcal{L}(\mathbb{R}) & \hookrightarrow & L^2(\mathbb{R}) & \hookrightarrow & \mathcal{L}'(\mathbb{R}) \\ \downarrow \mathcal{F} & & \downarrow \mathcal{F} & & \downarrow \mathcal{F} \\ \mathcal{L}(\mathbb{R}) & \hookrightarrow & L^2(\mathbb{R}) & \hookrightarrow & \mathcal{L}'(\mathbb{R}) \end{array}$$

commutes. The Plancherel Formula $\|f\|_2 = \|\hat{f}\|_2$ holds.

Proof : $\mathcal{F}(f)$ is defined as a distribution, so suffices to show 26.4
 $\|\mathcal{F}(f)\|_{L^2} < \infty$. Take $\psi = \hat{\varphi}$. Then

$$\hat{\psi} = \mathcal{F}^{-1}(\hat{\varphi}) = \int \bar{\varphi} e^{-i\xi x} dx = \mathcal{F}^{-1}(\bar{\varphi})$$

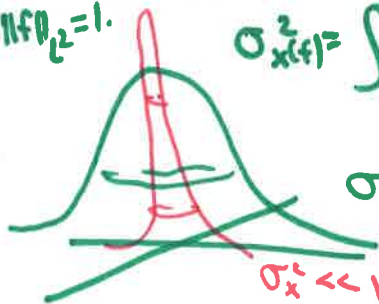
so $\hat{\psi} = \mathcal{F}\mathcal{F}^{-1}(\bar{\varphi}) = \bar{\varphi}$. Conclude

$$\int |\hat{\varphi}|^2 dx = \int |\varphi|^2 dx$$

□.

26.iv) The Uncertainty Principle

Let $f \in L^2$. Then the variance of $|f|$ is
 $\sigma_x^2(f) = \int_{-\infty}^{\infty} x^2 |f|^2 dx - \left(\int_{-\infty}^{\infty} x |f|^2 dx \right)^2$ ← mean



$$\sigma_x^2 \gg 1 \quad \sigma_y^2 = \sigma_x^2(\hat{f}).$$

Thm 26.14 (Heisenberg Uncertainty Principle)

$$\sigma_x \sigma_y \geq \frac{1}{2}.$$

Proof : may assume mean 0 by translation.

Let $\varphi = xf$ $\hat{\psi} = \xi \hat{f}$. so $\psi = -i \partial_x f$.

By Cauchy Schwartz,

$$|\langle \varphi, \psi \rangle|^2 \leq \left(\int x^2 f^2 \right) \left(\int \xi^2 \hat{f}^2 \right) = \sigma_x^2 \sigma_y^2.$$

But $|\langle \varphi, \psi \rangle|^2 \geq |\operatorname{Im} \langle \varphi, \psi \rangle|^2 \geq \left| \frac{\langle \varphi, \psi \rangle - \overline{\langle \varphi, \psi \rangle}}{2i} \right|^2$

And $-\langle \varphi, \psi \rangle + \langle \psi, \varphi \rangle = \int x \bar{f} (-i \partial_x f) - x f (i \partial_x \bar{f}) dx$
 $= \int x \bar{f} (-\partial_x f) + i \partial_x (xf) \bar{f} dx$
 $= \int x \bar{f} (-\partial_x f) + i f \bar{f} + i x \partial_x f \bar{f} dx = i \int |f|^2 dx = i$

$$= \left| \frac{i}{2} \right|^2 = \frac{1}{4}.$$

□.

Lecture 27 Fourier Transforms II: PDEs and Convolutions

(27.1)

Recall that on $L^2(S'; \mathbb{C})$, $\mathcal{F}$ turned Differential operators into multiplication.

Question 27.1: How do we extend the analysis on S' to $\mathbb{R}$?

How do we solve e.g.

$$\Delta u = f \in L^2(\mathbb{R})$$

$$(\Delta + 1)u = f$$

$$Lu = \sum \alpha_j(x) \partial_x^j u = f \in L^2(\mathbb{R})$$

or parabolic
$$\begin{cases} (\partial_t + \Delta_x) u(x, t) = g \\ u(x, 0) = f \end{cases}$$

and can we extend the results of e.g. elliptic regularity?

Observation 27.2

$$\begin{array}{ccc} L^{2,2}(\mathbb{R}) & \xleftrightarrow{\widehat{\mathcal{F}}} & L^2(\mathbb{R}, (1+|\xi|^2)^2 d\xi) \\ \downarrow L & & \downarrow \widehat{L} \\ L^2(\mathbb{R}) & \xleftrightarrow{\mathcal{F}} & L^2(\mathbb{R}) \end{array}$$

Note e.g. $\partial_x f \in L^2 \iff \int |\xi f|^2 d\xi = \int |f|^2 \underbrace{|\xi|^2}_{< \infty} d\xi < \infty$.

Thus to solve, one must study the "inverse operator"

$$\widehat{f} \longrightarrow \frac{1}{\alpha_0 + \alpha_1 \xi + \dots + \alpha_N \xi^N}$$

(for $\alpha_j \in \mathbb{C}$ constant coefficient).

Def 27.3: Let $K(z) := \mathcal{F}\left(\frac{1}{\alpha_0 + \dots + \alpha_N \xi^N}\right)$.

Then if $Lu = f$, $\Rightarrow u = \int_{\mathbb{R}} K(x-y) f(y) dy = \mathcal{F}^{-1}\left(\frac{1}{\alpha_0 + \dots + \alpha_N \xi^N} \cdot \widehat{f}\right)$.

$K(z)$ is called by various authors:

- 1) the Fundamental solution
- 2) the Green's function (of L)
- 3) the integral kernel
- 4) the Schwartz kernel.

Rem 27.4 :

$$\mathcal{F}(\delta)(\varphi) = \langle \delta, \mathcal{F}\varphi \rangle = \hat{\varphi}(0) = \int_{\mathbb{R}} \varphi(x) e^0 dx \\ = \int \langle \varphi(x), 1 \rangle dx$$

(27.2)

so $\mathcal{F}(\delta) = 1$. Thus

$$L K(\varepsilon) = \delta$$

is an equivalent definition, similar to as with the heat kernel.

Def 27.5 : The operator $u \mapsto K * u$ is called the
1) Green's operator, or (parametrix). It is a special
case of a convolution operator,

Convolution operators are a key aspect of the study of PDEs. See Math 1B/205.

27.ii: Young's Inequality for Convolutions

One fundamental result is

Thm 27.6 (Young's Convolution) $p, q, r \geq 1$.

Suppose $\frac{1}{p} + \frac{1}{q} = 1 + \frac{1}{r}$. Then $\exists C$ such that $f, g \in L^p(\mathbb{R}), L^q(\mathbb{R})$

$$\|f * g\|_r \leq C \|f\|_p \|g\|_q.$$

Hence $f * g \in L^r(\mathbb{R})$. The same holds on $\mathbb{R}^n$.

Proof :

$$f * g = \int_{\mathbb{R}} f(x-y)g(y)dy$$

$$\begin{aligned} |f * g| &\leq \int_{\mathbb{R}} |f(x-y)| |g(y)| dy \\ &\leq \int_{\mathbb{R}} |f(x-y)|^{1+p/r-p/r} |g(y)|^{1+q/r-q/r} dy \\ &\leq \int_{\mathbb{R}} (|f(x-y)|^p |g(y)|^q)^{1/r} |f(x-y)|^{1-p/r} |g(y)|^{1-q/r} dy \\ &\leq \left(\int_{\mathbb{R}} |f(x-y)|^p |g(y)|^q \right)^{1/r} \int_{\mathbb{R}} |f(x-y)|^{\frac{r-p}{r}} \| \frac{pr}{r-p} \| \\ &\quad \|g(y)\|^{\frac{r-q}{r}} \| \frac{qr}{r-q} \| \end{aligned}$$

Here, we used if $\frac{1}{p^*} + \frac{1}{q^*} + \frac{1}{r^*} = 1$

27.3

$$\int fgh \leq \|f\|_p \|g\|_q \|h\|_r$$

multi-Hölder, (induction) applied for $r^* = r$, $p^* = \frac{pr}{r-p}$, $q^* = \frac{qr}{r-q}$.

$$\leq \left(\int |f|^p |g|^q \right)^{1/r} \|f\|_p^{r-p} \|g\|_q^{r-q}$$

translation invariance of Lebesgue!

$$\|f+g\|_r^r = \int_{\mathbb{R}} |f+g|^r dx$$

$$\leq \int_{\mathbb{R}} \left[\int_{\mathbb{R}} |f(x-y)|^p |g(y)|^q dy \right] \|f\|_p^{r-p} \|g\|_q^{r-q} dx$$

$$\leq \|f\|_p^{r-p} \|g\|_q^{r-q} \int_{\mathbb{R}} \int_{\mathbb{R}} |g(y)|^q |f(x-y)|^p dx dy \quad (\text{Fubini})$$

$$\leq \|f\|_p^{r-p} \|g\|_q^{r-q} \int_{\mathbb{R}} g(y) \|f\|_p^p = \|f\|_p^{r-p+q} \|g\|_q^{r-q+q}$$

Ex 27.7 : On $\mathbb{R}^n$ $n \geq 3$

• $K(x-y)$ for Δ is $\frac{1}{|x-y|^{n-2}}$ so $\int |K(\frac{x}{r})| dv = \int \frac{1}{r^{n-2}} r^{n-1} dr$

$\Rightarrow$ the Green's function of Δ just barely fails to be L^1 !

• $K(x-y)$ for $\Delta + \beta^2 = \frac{e^{-\beta|x-y|}}{|x-y|^{n-2}}$ same (but only at 0).

Rem 27.7 : This is part of why studying integral kernels is difficult, cannot just apply Young.

27.iii) The Fourier transform of L^p

[27.4]

Recall $\mathcal{F}: L^2 \rightarrow L^2$ is an isometry.

Question 27.9: What is the Fourier transform of L^p for $1 \leq p < \infty$?

Ex 27.10 $\mathcal{F}(1) = \delta$ and $1 \in L^\infty$ but $\delta \in L^p$ for any p .

Thm (Hausdorff-Young) ^{27.11} If $1 \leq p \leq 2$, $\frac{1}{p} + \frac{1}{q} = 1$, then

$$\|\hat{f}\|_{L^q} \leq C \|f\|_{L^p}$$

Corollary 27.12: For $1 \leq p \leq 2$,

$$\mathcal{F}: L^p(\mathbb{R}) \longrightarrow L^q(\mathbb{R})$$

is continuous and injective. (It is not surjective) ^{this is extremely difficult.}

Proof: Suppose that $q = 2k$ for $k \in \mathbb{N}$, $p = \frac{2k}{2k-1} = 2, \frac{4}{3}, \frac{6}{5}, \dots$

$$\begin{aligned} \|\hat{f}\|_{L^q} &= \|\hat{f}\|_{L^{2k}} = \|(\hat{f})^k\|_{L^2}^{1/k} \\ &= \|\hat{f} \cdot \hat{f} \cdots \hat{f}\|_{L^2}^{1/k} \\ &= \|f * f * f * \cdots\|_{L^2}^{1/k} \end{aligned}$$

$$\text{If } k=2, \quad 1 + \frac{1}{2} = \frac{3}{2} = 2\left(\frac{3}{4}\right) = 2\left(\frac{1}{4}\right):$$

$$\leq \|f\|_{L^p} \|f\|_{L^p}^{1/2}$$

For general k , iterated Young convolution does same

o