Real Analysis Qualifying Exam Notes

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1 Inequalities and Equalities

AM-GM Inequality:

$$\sqrt{ab} \le \frac{a+b}{2}.$$

Reverse Triangle Inequality

$$|||x|| - ||y||| \le ||x - y||.$$

Chebyshev's Inequality

$$\mu(\lbrace x: |f(x)| > \alpha \rbrace) \le \left(\frac{\|f\|_p}{\alpha}\right)^p.$$

Holder's Inequality:

$$\frac{1}{p} + \frac{1}{q} = 1 \implies ||fg||_1 \le ||f||_p ||g||_q.$$

Application: For finite measure spaces,

$$1 \le p < q \le \infty \implies L^q \subset L^p \pmod{\ell^p \subset \ell^q}$$

Proof. Fix
$$p, q$$
, let $r = \frac{q}{p}$ and $s = \frac{r}{r-1}$ so $r^{-1} + s^{-1} = 1$. Then let $h = |f|^p$:
$$||f||_p^p = ||h \cdot 1||_1 \le ||1||_s ||h||_r = \mu(X)^{\frac{1}{s}} ||f||_q^{\frac{q}{r}} \implies ||f||_p \le \mu(X)^{\frac{1}{p} - \frac{1}{q}} ||f||_q.$$

Note: doesn't work for ℓ spaces, but just note that $\sum |x_n| < \infty \implies x_n < 1$ for large enough n, and thus $p < q \implies |x_n|^q \le |x_n|^q$.

Cauchy-Schwarz:

$$|\langle f, \; g \rangle| = \|fg\|_1 \leq \|f\|_2 \|g\|_2 \quad \text{with equality} \quad \Longleftrightarrow \; f = \lambda g.$$

Relates inner product to norm, and only happens to relate norms in L^1 .

 $\frac{Proof}{?}$

Minkowski's Inequality:

$$1 \le p < \infty \implies ||f + g||_p \le ||f||_p + ||g||_p$$

Note: does not handle $p = \infty$ case. Use to prove L^p is a normed space.

Young's Inequality:

$$\frac{1}{p} + \frac{1}{q} = \frac{1}{r} + 1 \implies ||f * g||_r \le ||f||_p ||g||_q.$$

Application: Some useful specific cases:

$$||f * g||_1 \le ||f||_1 ||g||_1$$

$$||f * g||_p \le ||f||_1 ||g||_p,$$

$$||f * g||_{\infty} \le ||f||_2 ||g||_2$$

$$||f * g||_{\infty} \le ||f||_p ||g||_q.$$

Bessel's Inequality:

For $x \in H$ a Hilbert space and $\{e_k\}$ an orthonormal sequence,

$$\sum_{k=1}^{\infty} |\langle x, e_k \rangle|^2 \le ||x||^2.$$

Note: this does not need to be a basis.

Parseval's Identity:

Equality in Bessel's inequality, attained when $\{e_k\}$ is a *basis*, i.e. it is complete, i.e. the span of its closure is all of H.

2 Basics

Useful Technique: $\lim f_n = \lim \sup f_n = \lim \inf f_n$ iff the limit exists, so $\lim \sup f_n \leq g \leq \lim \inf f_n$ implies that $g = \lim f$. Similarly, a limit does not exist iff $\lim \inf f_n > \lim \sup f_n$.

Lemma 2.1 (Convergent Sums Have Small Tails).

$$\sum a_n < \infty \implies a_n \longrightarrow 0 \text{ and } \sum_{k=N}^{\infty} \stackrel{N \longrightarrow \infty}{\longrightarrow} 0$$

Theorem 2.2 (Heine-Borel).

 $X \subseteq \mathbb{R}^n$ is compact $\iff X$ is closed and bounded.

Lemma 2.3 (Geometric Series).

$$\sum_{k=0}^{\infty} x^k = \frac{1}{1-x} \iff |x| < 1.$$

Corollary:
$$\sum_{k=0}^{\infty} \frac{1}{2^k} = 1.$$

Definition 2.3.1.

A set S is **nowhere dense** iff the closure of S has empty interior iff every interval contains a subinterval that does not intersect S.

Definition 2.3.2.

A set is **meager** if it is a *countable* union of nowhere dense sets.

Note that a *finite* union of nowhere dense is still nowhere dense.

Lemma 2.4.

The Cantor set is closed with empty interior.

Proof.

Its complement is a union of open intervals, and can't contain an interval since intervals have positive measure and $m(C_n)$ tends to zero.

Corollary: The Cantor set is nowhere dense.

Definition 2.4.1.

An F_{σ} set is a union of closed sets, and a G_{δ} set is an intersection of opens.

Mnemonic: "F" stands for ferme, which is "closed" in French, and σ corresponds to a "sum", i.e. a union.

Lemma 2.5.

Singleton sets in \mathbb{R} are closed, and thus \mathbb{Q} is an F_{σ} set.

Theorem 2.6 (Baire).

 \mathbb{R} is a **Baire space** (countable intersections of open, dense sets are still dense). Thus \mathbb{R} can not be written as a countable union of nowhere dense sets.

Lemma 2.7.

There is a function discontinuous precisely on \mathbb{Q} .

Proof .

 $f(x) = \frac{1}{n}$ if $x = r_n \in \mathbb{Q}$ is an enumeration of the rationals, and zero otherwise. The limit at every point is 0.

Lemma 2.8.

There do not exist functions that are discontinuous precisely on $\mathbb{R} \setminus \mathbb{Q}$.

Proof.

 D_f is always an F_σ set, which follows by considering the oscillation ω_f . $\omega_f(x) = 0 \iff f$ is continuous at x, and $D_f = \bigcup_x A_{\frac{1}{n}}$ where $A_\varepsilon = \{\omega_f \ge \varepsilon\}$ is closed.

Lemma 2.9.

Any nonempty set which is bounded from above (resp. below) has a well-defined supremum (resp. infimum).

3 Uniform Convergence

Theorem 3.1(Egorov).

Let $E \subseteq \mathbb{R}^n$ be measurable with m(E) > 0 and $\{f_k : E \longrightarrow \mathbb{R}\}$ be measurable functions such that

$$f(x) \coloneqq \lim_{k \to \infty} f_k(x) < \infty$$

exists almost everywhere.

Then $f_k \longrightarrow f$ almost uniformly, i.e.

 $\forall \varepsilon > 0, \ \exists F \subseteq E \text{ closed such that } m(E \setminus F) < \varepsilon \text{ and } f_k \xrightarrow{u} f \text{ on } F.$

Proposition 3.2.

The space X = C([0,1]), continuous functions $f : [0,1] \longrightarrow \mathbb{R}$, equipped with the norm $||f|| = \sup_{x \in [0,1]} |f(x)|$, is a **complete** metric space.

Proof.

- 1. Let $\{f_k\}$ be Cauchy in X.
- 2. Define a candidate limit using pointwise convergence: Fix an x; since

$$|f_k(x) - f_i(x)| \le ||f_k - f_k|| \longrightarrow 0$$

the sequence $\{f_k(x)\}\$ is Cauchy in \mathbb{R} . So define $f(x) := \lim_{k \to \infty} f_k(x)$.

3. Show that $||f_k - f|| \longrightarrow 0$:

$$|f_k(x) - f_j(x)| < \varepsilon \ \forall x \implies \lim_j |f_k(x) - f_j(x)| < \varepsilon \ \forall x$$

Alternatively, $||f_k - f|| \le ||f_k - f_N|| + ||f_N - f_j||$, where N, j can be chosen large enough to bound each term by $\varepsilon/2$.

4. Show that $f \in X$:

The uniform limit of continuous functions is continuous. (Note: in other cases, you may need to show the limit is bounded, or has bounded derivative, or whatever other conditions define X.)

Lemma 3.3.

Metric spaces are compact iff they are sequentially compact, (i.e. every sequence has a convergent subsequence).

Proposition 3.4.

The unit ball in C([0,1]) with the sup norm is not compact.

Proof.

Take $f_k(x) = x^n$, which converges to a dirac delta at 1. The limit is not continuous, so no subsequence can converge.

Lemma 3.5.

A uniform limit of continuous functions is continuous.

Lemma 3.6 (Testing Uniform Convergence).

 $f_n \longrightarrow f$ uniformly iff there exists an M_n such that $||f_n - f||_{\infty} \leq M_n \longrightarrow 0$.

Negating: find an x which depends on n for which the norm is bounded below.

Useful Technique: If f_n has a global maximum (computed using f'_n and the first derivative test) $M_n \longrightarrow 0$, then $f_n \longrightarrow 0$ uniformly.

Lemma 3.7 (Baby Commuting Limits with Integrals).

If $f_n \longrightarrow f$ uniformly, then $\int f_n = \int f$.

Lemma 3.8 (Uniform Convergence and Derivatives).

If $f'_n \longrightarrow g$ uniformly for some g and $f_n \longrightarrow f$ pointwise (or at least at one point), then g = f'.

Lemma 3.9 (Uniform Convergence of Series).

If $f_n(x) \leq M_n$ for a fixed x where $\sum M_n < \infty$, then the series $f(x) = \sum f_n(x)$ converges pointwise.

Lemma 3.10.

If $\sum f_n$ converges then $f_n \longrightarrow 0$ uniformly.

Useful Technique: For a fixed x, if $f = \sum_{n} f_n$ converges uniformly on some $B_r(x)$ and each f_n is continuous at x, then f is also continuous at x.

Lemma 3.11 (M-test for Series).

If $|f_n(x)| \leq M_n$ which does not depend on x, then $\sum f_n$ converges uniformly.

Lemma 3.12(p-tests).

Let n be a fixed dimension and set $B = \{x \in \mathbb{R}^n \mid ||x|| \le 1\}.$

$$\sum_{n} \frac{1}{n^{p}} < \infty \iff p > 1$$

$$\int_{\varepsilon}^{\infty} \frac{1}{x^{p}} < \infty \iff p > 1$$

$$\int_{0}^{1} \frac{1}{x^{p}} < \infty \iff p < 1$$

$$\int_{B} \frac{1}{|x|^{p}} < \infty \iff p < n$$

$$\int_{B^{c}} \frac{1}{|x|^{p}} < \infty \iff p > n$$

4 Measure Theory

Useful Technique: $s = \inf \{x \in X\} \implies \text{for every } \varepsilon \text{ there is an } x \in X \text{ such that } x \leq s + \varepsilon.$

Useful Techniques: Always consider bounded sets, and if E is unbounded write $E = \bigcup_n B_n(0) \cap E$ and use countable subadditivity or continuity of measure.

Lemma 4.1.

Every open subset of \mathbb{R} (resp \mathbb{R}^n) can be written as a unique countable union of disjoint (resp. almost disjoint) intervals (resp. cubes).

Definition 4.1.1.

The outer measure of a set is given by

$$m_*(E) = \inf_{\substack{\{Q_i\} \rightrightarrows E \ \text{closed cubes}}} \sum |Q_i|.$$

Lemma 4.2 (Properties of Outer Measure).

• Montonicity: $E \subseteq F \implies m_*(E) \le m_*(F)$.

- Countable Subadditivity: m_{*}(∪ E_i) ≤ ∑ m_{*}(E_i).
 Approximation: For all E there exists a G ⊇ E such that m_{*}(G) ≤ m_{*}(E) + ε.
- Disjoint^a Additivity: $m_*(A \coprod B) = m_*(A) + m_*(B)$.

Lemma 4.3 (Subtraction of Measure).

$$m(A) = m(B) + m(C), m(C) < \infty \implies m(A) - m(C) = m(B).$$

Lemma 4.4(Continuity of Measure).

$$E_i \nearrow E \implies m(E_i) \longrightarrow m(E)$$

 $m(E_1) < \infty \text{ and } E_i \searrow E \implies m(E_i) \longrightarrow m(E).$

- 1. Break into disjoint annuli $A_2 = E_2 \setminus E_1$, etc then apply countable disjoint additivity to $E = \prod A_i$.
 - 2. Use $E_1 = (\prod E_j \setminus E_{j+1}) \prod (\bigcap E_j)$, taking measures yields a telescoping sum, and use countable disjoint additivity.

Lemma 4.5.

Lebesgue measure is translation and dilation invariant.

Proof.

Obvious for cubes; if $Q_i \rightrightarrows E$ then $Q_i + k \rightrightarrows E + k$, etc.

Theorem 4.6 (Non-Measurable Sets).

There is a non-measurable set.

Proof.

- Use AOC to choose one representative from every coset of \mathbb{R}/\mathbb{Q} on [0,1), which is countable, and assemble them into a set N
- Enumerate the rationals in [0,1] as q_j , and define $N_j = N + q_j$. These intersect trivially.
- Define $M := \coprod N_j$, then $[0,1) \subseteq M \subseteq [-1,2)$, so the measure must be between 1 and 3. By translation invariance, $m(N_j) = m(N)$, and disjoint additivity forces m(M) = 0, a contradiction.

4 MEASURE THEORY

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^aThis holds for outer measure **iff** dist(A, B) > 0.

Proposition 4.7 (Borel Characterization of Measurable Sets).

If E is Lebesgue measurable, then $E = H \coprod N$ where $H \in F_{\sigma}$ and N is null.

Useful technique: F_{σ} sets are Borel, so establish something for Borel sets and use this to extend it to Lebesgue.

Proof.

For every $\frac{1}{n}$ there exists a closed set $K_n \subset E$ such that $m(E \setminus K_n) \leq \frac{1}{n}$. Take $K = \bigcup K_n$, wlog $K_n \nearrow K$ so $m(K) = \lim m(K_n) = m(E)$. Take $N \coloneqq E \setminus K$, then m(N) = 0.

Definition 4.7.1.

$$\limsup_{n} A_{n} := \bigcap_{n} \bigcup_{j \ge n} A_{j} = \left\{ x \mid x \in A_{n} \text{ for inf. many } n \right\}$$
$$\liminf_{n} A_{n} := \bigcup_{n} \bigcap_{j \ge n} A_{j} = \left\{ x \mid x \in A_{n} \text{ for all except fin. many } n \right\}$$

Lemma 4.8.

If A_n are all measurable, $\limsup A_n$ and $\liminf A_n$ are measurable.

Proof.

Measurable sets form a sigma algebra, and these are expressed as countable unions/intersections of measurable sets.

Theorem 4.9 (Borel-Cantelli).

Let $\{E_k\}$ be a countable collection of measurable sets. Then

 $\sum_{k} m(E_k) < \infty \implies \text{ almost every } x \in \mathbb{R} \text{ is in at most finitely many } E_k.$

Proof.

- If $E = \limsup_{j \to \infty} E_j$ with $\sum_{j \to \infty} m(E_j) < \infty$ then m(E) = 0.
- If E_j are measurable, then $\limsup_{i} E_j$ is measurable.
- If $\sum_{j} m(E_{j}) < \infty$, then $\sum_{j=N}^{\infty} m(E_{j}) \stackrel{N \longrightarrow \infty}{\longrightarrow} 0$ as the tail of a convergent sequence. $E = \limsup_{j} E_{j} = \bigcap_{k=1}^{\infty} \bigcup_{j=k}^{\infty} E_{j} \implies E \subseteq \bigcup_{j=k}^{\infty} \text{ for all } k$

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$$E \subset \bigcup_{j=k}^{\infty} \implies m(E) \leq \sum_{j=k}^{\infty} m(E_j) \stackrel{k \longrightarrow \infty}{\longrightarrow} 0.$$

Lemma 4.10.

- Characteristic functions are measurable
- If f_n are measurable, so are $|f_n|$, $\limsup f_n$, $\liminf f_n$, $\lim f_n$,
- Sums and differences of measurable functions are measurable,
- Cones F(x,y) = f(x) are measurable,
- Compositions $f \circ T$ for T a linear transformation are measurable,
- "Convolution-ish" transformations $(x,y) \mapsto f(x-y)$ are measurable

Proof (Convolution).

Take the cone on f to get F(x,y) = f(x), then compose F with the linear transformation T = [1, -1; 1, 0].

5 Integration

Definition 5.0.1.

 $f \in L^+$ iff f is measurable and non-negative.

Useful techniques:

- Break integration domain up into disjoint annuli.
- Break real integrals up into x < 1 and x > 1.

Definition 5.0.2.

A measurable function is integrable iff $||f||_1 < \infty$.

Useful facts about C_c functions:

- Bounded almost everywhere
- Uniformly continuous

5.1 Convergence Theorems

Theorem 5.1 (Monotone Convergence).

If $f_n \in L^+$ and $f_n \nearrow f$ a.e., then

$$\lim \int f_n = \int \lim f_n = \int f \quad \text{i.e. } \int f_n \longrightarrow \int f.$$

Needs to be positive and increasing.

Theorem 5.2(Dominated Convergence).

If $f_n \in L^1$ and $f_n \longrightarrow f$ a.e. with $|f_n| \leq g$ for some $g \in L^1$, then

$$\lim \int f_n = \int \lim f_n = \int f$$
 i.e. $\int f_n \longrightarrow \int f$,

and more generally,

$$\int |f_n - f| \longrightarrow 0.$$

Positivity *not* needed.

Generalized DCT: can relax $|f_n| < g$ to $|f_n| < g_n \longrightarrow g \in L^1$.

Lemma 5.3.

If $f \in L^1$, then

$$\int |f_n - f| \longrightarrow 0 \iff \int |f_n| \longrightarrow |f|.$$

Proof

Let $g_n = |f_n| - |f_n - f|$, then $g_n \longrightarrow |f|$ and

$$|g_n| = ||f_n| - |f_n - f|| \le |f_n - (f_n - f)| = |f| \in L^1,$$

so the DCT applies to g_n and

$$||f_n - f||_1 = \int |f_n - f| + |f_n| - |f_n| = \int |f_n| - g_n$$

$$\longrightarrow_{DCT} \lim \int |f_n| - \int |f|.$$

Fatou's Lemma If $f_n \in L^+$, then

$$\int \liminf_{n} f_n \le \liminf_{n} \int f_n$$
$$\lim \sup_{n} \int f_n \le \int \limsup_{n} f_n.$$

Only need positivity.

Theorem 5.4 (Tonelli).

For f(x,y) non-negative and measurable, for almost every $x \in \mathbb{R}^n$,

- $f_x(y)$ is a **measurable** function
- $F(x) = \int f(x,y) dy$ is a **measurable** function,
- For E measurable, the slices $E_x := \{y \mid (x,y) \in E\}$ are measurable.

• $\int f = \int \int F$, i.e. any iterated integral is equal to the original.

Theorem 5.5 (Fubini).

For f(x,y) integrable, for almost every $x \in \mathbb{R}^n$,

- $f_x(y)$ is an **integrable** function
- $F(x) = \int f(x,y) dy$ is an **integrable** function,
- For E measurable, the slices $E_x := \{y \mid (x,y) \in E\}$ are measurable.
- $\int f = \int \int f(x,y)$, i.e. any iterated integral is equal to the original

Theorem 5.6 (Fubini/Tonelli).

If any iterated integral is **absolutely integrable**, i.e. $\int \int |f(x,y)| < \infty$, then f is integrable and $\int f$ equals any iterated integral.

Useful Technique: Differentiating under the integral.

If
$$\left| \frac{\partial}{\partial t} f(x,t) \right| \leq g(x) \in L^1$$
, then letting $F(t) = \int f(x,t) \ dt$,

$$\frac{\partial}{\partial t} F(t) \coloneqq \lim_{h \to 0} \int \frac{f(x, t+h) - f(x, t)}{h} dx$$
$$\stackrel{\text{DCT}}{=} \int \frac{\partial}{\partial t} f(x, t) \ dx.$$

To justify passing the limit, let $h_k \longrightarrow 0$ be any sequence and define

$$f_k(x,t) = \frac{f(x,t+h_k) - f(x,t)}{h_k},$$

so
$$f_k \stackrel{\text{pointwise}}{\longrightarrow} \frac{\partial}{\partial t} f$$
.

Apply the MVT to f_k to get $f_k(x,t) = f_k(\xi,t)$ for some $\xi \in [0,h_k]$, and show that $f_k(\xi,t) \in L_1$.

Proposition 5.7(Swapping Sum and Integral).

If f_n are non-negative and $\sum \int |f|_n < \infty$, then $\sum \int f_n = \int \sum f_n$.

Proof

MCT. Let $F_N = \sum_{n=1}^{N} f_n$ be a finite partial sum; then there are simple functions $\phi_n \nearrow f_n$ and so $\sum_{n=1}^{N} \phi_n \nearrow F_N$, so apply MCT.

Lemma 5.8.

If $f_k \in L^1$ and $\sum ||f_k||_1 < \infty$ then $\sum f_k$ converges almost everywhere and in L^1 .

Proof.

Define
$$F_N = \sum_{k=1}^{N} f_k$$
 and $F = \lim_{k \to \infty} F_k$, then $||F_N||_1 \le \sum_{k=1}^{N} ||f_k|| < \infty$ so $F \in L^1$ and $||F_N - F||_1 \longrightarrow 0$ so the sum converges in L^1 . Almost everywhere convergence: ?

5.2 L^1 Facts

Lemma 5.9 (Translation Invariance).

The Lebesgue integral is translation invariant, i.e. $\int f(x) dx = \int f(x+h) dx$ for any h.

Proof .

- For characteristic functions, $\int_E f(x+h) = \int_{E+h} f(x) = m(E+h) = m(E) = \int_E f$ by translation invariance of measure.
- So this also holds for simple functions by linearity
- For $f \in L^+$, choose $\phi_n \nearrow f$ so $\int \phi_n \longrightarrow \int f$.
- Similarly, $\tau_h \phi_n \nearrow \tau_h f$ so $\int \tau_h f \longrightarrow \int f$
- Finally $\left\{ \int \tau_h \phi \right\} = \left\{ \int \phi \right\}$ by step 1, and the suprema are equal by uniqueness of limits.

 ${\bf Lemma~5.10} (Integrals~Distribute~Over~Disjoint~Sets).$

If
$$X \subseteq A \bigcup B$$
, then $\int_X f \leq \int_A f + \int_{A^c} f$ with equality iff $X = A \coprod B$.

Lemma 5.11 (Unif Cts L1 Functions Decay Rapidly).

If
$$f \in L^1$$
 and f is uniformly continuous, then $f(x) \stackrel{|x| \longrightarrow \infty}{\longrightarrow} 0$.

Doesn't hold for general L^1 functions, take any train of triangles with height 1 and summable areas.

Lemma 5.12(L1 Functions Have Small Tails).

If
$$f \in L^1$$
, then for every ε there exists a radius R such that if $A = B_R(0)^c$, then $\int_A |f| < \varepsilon$.

Proof.

Approximate with compactly supported functions. Take $g \xrightarrow{L_1} f$ with $g \in C_c$, then choose N

large enough so that g=0 on $E:=B_N(0)^c$, then $\int_E |f| \le \int_E |f-g| + \int_E |g|$.

Lemma 5.13(L1 Functions Have Absolutely Continuity). $m(E) \longrightarrow 0 \implies \int_E f \longrightarrow 0.$

Proof.

Approximate with compactly supported functions. Take $g \xrightarrow{L_1} f$, then $g \leq M$ so $\int_E f \leq \int_E f - g + \int_E g \longrightarrow 0 + M \cdot m(E) \longrightarrow 0$.

Lemma 5.14(L1 Functions Are Finite a.e.). If $f \in L^1$, then $m(\{f(x) = \infty\}) = 0$.

Proof.

Idea: Split up domain Let $A = \{f(x) = \infty\}$, then $\infty > \int f = \int_A f + \int_{A^c} f = \infty \cdot m(A) + \int_{A^c} f \implies m(X) = 0.$

Lemma (Continuity in L^1): $\|\tau_h f - f\|_1 \longrightarrow 0$ as $h \longrightarrow 0$.

Proof: Approximate with compactly supported functions. Take $g \xrightarrow{L_1} f$ with $g \in C_c$.

$$\int f(x+h) - f(x) \le \int f(x+h) - g(x+h) + \int g(x+h) - g(x) + \int g(x) - f(x)$$

$$\longrightarrow 2\varepsilon + \int g(x+h) - g(x)$$

$$= \int_{K} g(x+h) - g(x) + \int_{K^{c}} g(x+h) - g(x) \longrightarrow 0,$$

which follows because we can enlarge the support of g to K where the integrand is zero on K^c , then apply uniform continuity on K.

Theorem (Integration by Parts, Special Case):

$$F(x) := \int_0^x f(y)dy \quad \text{and} \quad G(x) := \int_0^x g(y)dy$$

$$\implies \int_0^1 F(x)g(x)dx = F(1)G(1) - \int_0^1 f(x)G(x)dx.$$

Proof: Fubini-Tonelli, and sketch region to change integration bounds.

Theorem (Lebesgue Density):

$$A_h(f)(x) := \frac{1}{2h} \int_{x-h}^{x+h} f(y) dy \implies ||A_h(f) - f|| \stackrel{h \longrightarrow 0}{\longrightarrow} 0.$$

Proof: Fubini-Tonelli, and sketch region to change integration bounds, and continuity in L^1 .

5.3 L^p Spaces

Lemma: The following are dense subspaces of $L^2([0,1])$:

- Simple functions
- Step functions
- $C_0([0,1])$
- Smoothly differentiable functions $C_0^{\infty}([0,1])$
- Smooth compactly supported functions C_c^{∞}

Dual Spaces: In general, $(L^p)^{\vee} \cong L^q$

- \bullet For qual, supposed to know the p=1 case, i.e. $(L^1)^\vee \cong L^\infty$
 - For the analogous $p=\infty$ case: $L^1\subset (L^\infty)^\vee$, since the isometric mapping is always injective, but *never* surjective. So this containment is always proper (requires Hahn-Banach Theorem).
- The p=2 case: Easy by the Riesz Representation for Hilbert spaces.

6 Fourier Series and Convolution

Definition (Convolution)

$$f * g(x) = \int f(x - y)g(y)dy.$$

Definition (Dilation)

$$\phi_t(x) = t^{-n}\phi\left(t^{-1}x\right).$$

Definition (The Fourier Transform):

$$\widehat{f}(\xi) = \int f(x) \ e^{2\pi i x \cdot \xi} \ dx.$$

Lemma: $\hat{f} = \hat{g} \implies f = g$ almost everywhere.

Lemma (Riemann-Lebesgue)

$$f \in L^1 \implies \widehat{f}(\xi) \to 0 \text{ as } |\xi| \to \infty.$$

Motto: Fourier transforms decay.

Lemma: If $f \in L^1$, then \hat{f} is continuous and bounded.

Proof:
$$|\widehat{f}| \leq \int |f| \cdot |e^{\cdots}| \leq ||f||_1$$
, and the DCT shows that $|\widehat{f}(\xi_n) - \widehat{f}(\xi)| \longrightarrow 0$.

Todo: search qual alerts.

7 Exam 2 (Practice)

Link to PDF File

Proving uniform continuity: show

$$||f - \tau_h f||_1 \xrightarrow{h \longrightarrow 0} 0$$

Notation: C_0 is the set of functions that vanish at infinity.

7.1 1: Fubini-Tonelli

Theorem (Fubini):

Suppose

- $f: \mathbb{R}^{n_1+n_2} \longrightarrow \mathbb{R}$ is measurable on its domain
- \bullet f is non-negative

Then for almost every $x \in \mathbb{R}^{n_1}$,

1. Every slice

$$f_x: \mathbb{R}^{n_2} \longrightarrow \mathbb{R}$$

 $y \mapsto f(x, y)$

is measurable on \mathbb{R}^{n_2} .

2. The function

$$F: \mathbb{R}^{n_1} \longrightarrow \mathbb{R}$$
$$x \mapsto \int_{\mathbb{R}^{n_2}} f_x(y) \ dy$$

is measurable on \mathbb{R}^{n_1}

3. The function

$$G(y) = \int_{\mathbb{R}^n} F(x) \ dx$$

is measurable and

$$G(y) = \int_{\mathbb{R}^n} f = \int_{\mathbb{R}^{n_1}} \int_{\mathbb{R}^{n_2}} f(x, y) \ dy \ dx$$

for any iterated version of this integral.

Corollary (Measurable Slices):

Let E be a measurable subset of \mathbb{R}^n . Then

- For almost every $x \in \mathbb{R}^{n_1}$, the slice $E_x := \{ y \in \mathbb{R}^{n_2} \mid (x,y) \in E \}$ is measurable in \mathbb{R}^{n_2} .
- The function

$$F: \mathbb{R}^{n_1} \longrightarrow \mathbb{R}$$

$$x \mapsto m(E_x) = \int_{\mathbb{R}^{n_2}} \chi_{E_x} \ dy$$

is measurable and

$$m(E) = \int_{\mathbb{R}^{n_1}} m(E_x) \ dx = \int_{\mathbb{R}^{n_1}} \int_{\mathbb{R}^{n_2}} \chi_{E_x} \ dy \ dx$$

 \Longrightarrow :

- Let f be measurable on \mathbb{R}^n .
- Then the cylinders F(x,y) = f(x) and G(x,y) = f(y) are both measurable on \mathbb{R}^{n+1} .
- Write $\mathcal{A} = \{G \leq F\} \cap \{G \geq 0\}$; both are measurable.

- Let A be measurable in \mathbb{R}^{n+1} .
- Define $A_x = \{y \in \mathbb{R} \mid (x,y) \in \mathcal{A}\}$, then $m(A_x) = f(x)$. By the corollary, A_x is measurable set, $x \mapsto A_x$ is a measurable function, and m(A) = f(x). $\int f(x) dx$.
- Then explicitly, $f(x) = \chi_A$, which makes f a measurable function.

7.1.1 b

- Define $A_y = \{x \in \mathbb{R}^n \mid (x, y) \in A\}$, and notice that $A_y = \{x \in \mathbb{R}^n \mid 0 \le y \le f(x)\}$.
- By the corollary, A_y is measurable and

$$m(\mathcal{A}) = \int m(\mathcal{A}_y) \, dy = \int_0^y m\left(\left\{x \in \mathbb{R}^n \ni f(x) \ge y\right\}\right) dy$$

7.2 2: Convolutions and the Fourier Transform

7.2.1 a

Definition (Convolution):

$$(f * g)(x) = \int_{\mathbb{R}^n} f(x - y)g(y) \ dy.$$

Facts:

- $\bullet \ f,g \in L^1 \implies f * g \in L^1$
- $f \in L^1, g \leq M \implies f * g \leq M'$ and is uniformly continuous.

- $f, g \in L^1, f < M, g < M' \implies f * g \xrightarrow{x \to \infty} 0.2$
- $||f * g||_1 \le ||f||_1 ||g||_1$ $f \in L^1, g' \text{ exists}, \frac{\partial g}{\partial x_i} \text{ all bounded } \implies \frac{\partial}{\partial x_i} (f * g) = f * \frac{\partial g}{\partial x_i}$
- $f, g \in C_c^{\infty} \implies f * g \in C^{\infty} \text{ and } f * g \xrightarrow{x \to \infty} 0.$

7.2.2 b

Definition (Approximation to the Identity):

$$\phi(x) = e^{-\pi x^2}$$
$$\phi_t(x) = t^{-n}\phi(\frac{x}{t}).$$

Facts:

- $\int \phi = \int \phi_t = 1$ f bounded and uniformly continuous $\implies f * \phi_t \rightrightarrows f$

Theorem (Norm Convergence of Approximate Identities):

$$||f * \phi_t - f||_1 \xrightarrow{t \longrightarrow 0} 0.$$

Proof:

$$\begin{split} \|f-f*\phi_t\|_1 &= \int f(x) - \int f(x-y)\phi_t(y) \; dy dx \\ &= \int f(x) \int \phi_t(y) \; dy - \int f(x-y)\phi_t(y) \; dy dx \\ &= \int \int \phi_t(y)[f(x) - f(x-y)] \; dy dx \\ &=_{FT} \int \int \phi_t(y)[f(x) - f(x-y)] \; dx dy \\ &= \int \phi_t(y) \int f(x) - f(x-y) \; dx dy \\ &= \int \phi_t(y) \|f - \tau_y f\|_1 dy \\ &= \int_{y < \delta} \phi_t(y) \|f - \tau_y f\|_1 dy + \int_{y \ge \delta} \phi_t(y) \|f - \tau_y f\|_1 dy \\ &\leq \int_{y < \delta} \phi_t(y) \varepsilon + \int_{y \ge \delta} \phi_t(y) \left(\|f\|_1 + \|\tau_y f\|_1\right) dy \quad \text{by continuity in } L^1 \\ &\leq \varepsilon + 2\|f\|_1 \int_{y \ge \delta} \phi_t(y) dy \\ &\leq \varepsilon + 2\|f\|_1 \varepsilon \quad \text{since } \phi_t \text{ has small tails} \\ &\to 0 \blacksquare. \end{split}$$

Theorem (Convolutions Vanish at Infinity)

$$f, g \in L^1$$
 and bounded $\Longrightarrow \lim_{|x| \to \infty} (f * g)(x) = 0.$

Proof:

• Choose $M \geq f, g$.

• By small tails, choose N such that $\int_{B_n^c} |f|, \int_{B_n^c} |g| < \varepsilon$

• Note

$$|f * g| \le \int |f(x-y)| |g(y)| dy := I$$

• Use $|x| \le |x - y| + |y|$, take $|x| \ge 2N$ so either

 $|x-y| \ge N \implies I \le \int_{\{x-y \ge N\}} |f(x-y)| M \ dy \le \varepsilon M \longrightarrow 0$

 $|y| \ge N \implies I \le \int_{\{y \ge N\}} M|g(y)| \ dy \le M\varepsilon \longrightarrow 0$

7.2.3 c

Definition (The Fourier Transform):

$$\widehat{f}(\xi) = \int f(x)e^{-2\pi ix\cdot\xi} dx.$$

Facts:

• Riemann-Lebesgue lemma: \hat{f} vanishes at infinity • $f \in L^1 \implies \hat{f}$ is bounded and uniformly continuous • $f, \hat{f} \in L^1 \implies f$ is bounded, uniformly continuous, and vanishes at infinity • $f \in L^1$ and $\hat{f} = 0$ almost everywhere $\implies f = 0$ almost everywhere.

Theorem (Fourier Inversion):

$$f(x) = \int_{\mathbb{R}^n} \widehat{f}(x)e^{2\pi ix\cdot\xi}d\xi.$$

Proof: Idea: Fubini-Tonelli doesn't work directly, so introduce a convergence factor, take limits, and use uniqueness of limits.

Use the following facts:

• $f, g \in L^1 \Longrightarrow \int \widehat{f}g = \int f\widehat{g}.$ • $g(x) \coloneqq e^{-\pi|t|^2} \Longrightarrow \widehat{g}(\xi) = g(\xi)$ • $g_t(x) = t^{-n}g(x/t) = t^{-n}e^{-\pi|x|^2/t^2}$

•
$$\widehat{g}_t(x) = g(tx) = e^{-\pi t^2 |x|^2}$$

• $\phi(\xi) := e^{2\pi i x \cdot \xi} \ \widehat{g}_t(\xi)$

•
$$\phi(\xi) := e^{2\pi i x \cdot \xi} \ \widehat{g}_t(\xi)$$

•
$$\widehat{\phi}(\xi) = \mathcal{F}(\widehat{g}_t(\xi - x)) = g_t(x - \xi)$$

• $\lim_{t \to 0} \phi(\xi) = e^{2\pi i x \cdot \xi}$

•
$$\lim_{\xi \to 0} \phi(\xi) = e^{2\pi i x \cdot \xi}$$

Take the modified integral:

$$I_{t}(x) = \int \widehat{f}(\xi) e^{2\pi i x \cdot \xi} e^{-\pi t^{2} |\xi|^{2}}$$

$$= \int \widehat{f}(\xi) \phi(\xi)$$

$$= \int f(\xi) \widehat{\phi}(\xi)$$

$$= \int f(\xi) \widehat{g}(\xi - x)$$

$$= \int f(\xi) g_{t}(x - \xi) d\xi$$

$$= \int f(y - x) g_{t}(y) dy \quad (\xi = y - x)$$

$$= (f * g_{t})$$

$$\longrightarrow f \text{ in } L^{1} \text{ as } t \longrightarrow 0,$$

but we also have

$$\lim_{t \to 0} I_t(x) = \lim_{t \to 0} \int \widehat{f}(\xi) \ e^{2\pi i x \cdot \xi} \ e^{-\pi t^2 |\xi|^2}$$

$$= \lim_{t \to 0} \int \widehat{f}(\xi) \phi(\xi)$$

$$=_{DCT} \int \widehat{f}(\xi) \lim_{t \to 0} \phi(\xi)$$

$$= \int \widehat{f}(\xi) \ e^{2\pi i x \cdot \xi}$$

So

$$I_t(x) \longrightarrow \int \widehat{f}(\xi) \ e^{2\pi i x \cdot \xi} \ \text{ pointwise and } \|I_t(x) - f(x)\|_1 \longrightarrow 0.$$

So there is a subsequence I_{t_n} such that $I_{t_n}(x) \longrightarrow f(x)$ almost everywhere, so $f(x) = \int \hat{f}(\xi) e^{2\pi i x \cdot \xi}$ almost everywhere by uniqueness of limits.

7.3 3: Hilbert Spaces

7.3.1 a

Theorem (Bessel's Inequality):

$$\left\| x - \sum_{n=1}^{N} \langle x, u_n \rangle u_n \right\|^2 = \|x\|^2 - \sum_{n=1}^{N} |\langle x, u_n \rangle|^2$$

and thus

$$\sum_{n=1}^{\infty} |\langle x, u_n \rangle|^2 \le ||x||^2$$

Proof: Let
$$S_N = \sum_{n=1}^N \langle x, u_n \rangle u_n$$

$$||x - S_N||^2 = \langle x - S_n, x - S_N \rangle$$

$$= ||x||^2 + ||S_N||^2 - 2\Re \langle x, S_N \rangle$$

$$= ||x||^2 + ||S_N||^2 - 2\Re \left\langle x, \sum_{n=1}^N \langle x, u_n \rangle u_n \right\rangle$$

$$= ||x||^2 + ||S_N||^2 - 2\Re \sum_{n=1}^N \langle x, \langle x, u_n \rangle u_n \rangle$$

$$= ||x||^2 + ||S_N||^2 - 2\Re \sum_{n=1}^N \overline{\langle x, u_n \rangle} \langle x, u_n \rangle$$

$$= ||x||^2 + ||\sum_{n=1}^N \langle x, u_n \rangle u_n ||^2 - 2\sum_{n=1}^N |\langle x, u_n \rangle|^2$$

$$= ||x||^2 + \sum_{n=1}^N |\langle x, u_n \rangle|^2 - 2\sum_{n=1}^N |\langle x, u_n \rangle|^2$$

And by continuity of the norm and inner product, we have

$$\lim_{N \to \infty} \|x - S_N\|^2 = \lim_{N \to \infty} \|x\|^2 - \sum_{n=1}^N |\langle x, u_n \rangle|^2$$

$$\implies \left\| x - \lim_{N \to \infty} S_N \right\|^2 = \|x\|^2 - \lim_{N \to \infty} \sum_{n=1}^N |\langle x, u_n \rangle|^2$$

$$\implies \left\| x - \sum_{n=1}^\infty \langle x, u_n \rangle u_n \right\|^2 = \|x\|^2 - \sum_{n=1}^\infty |\langle x, u_n \rangle|^2.$$

 $= ||x||^2 - \sum_{n=1}^{N} |\langle x, u_n \rangle|^2.$

Then noting that $0 \le ||x - S_N||^2$, we have

$$0 \le ||x||^2 - \sum_{n=1}^{\infty} |\langle x, u_n \rangle|^2$$

$$\implies \sum_{n=1}^{\infty} |\langle x, u_n \rangle|^2 \le ||x||^2 \blacksquare.$$

7.3.2 b

• Take $\{a_n\} \in \ell^2$, then note that $\sum |a_n|^2 < \infty \implies$ the tails vanish.

• Define
$$x = \lim_{N \to \infty} S_N$$
 where $S_N = \sum_{k=1}^N a_k u_k$

• $\{S_N\}$ is Cauchy and H is complete, so $x \in H$.

- By construction, $\langle x, u_n \rangle = \left\langle \sum_k a_k u_k, u_n \right\rangle = \sum_k a_k \langle u_k, u_n \rangle = a_n$ since the u_k are all orthogonal.
- $||x||^2 = \left\|\sum_k a_k u_k\right\|^2 = \sum_k ||a_k u_k||^2 = \sum_k |a_k|^2$ by Pythagoras since the u_k are normal.

7.3.3 c

Let x and u_n be arbitrary. Then

$$\left\langle x - \sum_{k=1}^{\infty} \langle x, u_k \rangle u_k, u_n \right\rangle = \langle x, u_n \rangle - \left\langle \sum_{k=1}^{\infty} \langle x, u_k \rangle u_k, u_n \right\rangle$$

$$= \langle x, u_n \rangle - \sum_{k=1}^{\infty} \langle \langle x, u_k \rangle u_k, u_n \rangle$$

$$= \langle x, u_n \rangle - \sum_{k=1}^{\infty} \langle x, u_k \rangle \langle u_k, u_n \rangle$$

$$= \langle x, u_n \rangle - \langle x, u_n \rangle = 0$$

$$\implies x - \sum_{k=1}^{\infty} \langle x, u_k \rangle u_k = 0 \quad \text{by completeness.}$$

So

$$x = \sum_{k=1}^{\infty} \langle x, u_k \rangle u_k \implies ||x||^2 = \sum_{k=1}^{\infty} |\langle x, u_k \rangle|^2. \blacksquare$$

7.4 4: Lp Spaces

p-test for integrals:

$$\int_{0}^{1} x^{-p} < \infty \iff p < 1$$
$$\int_{1}^{\infty} x^{-p} < \infty \iff p > 1.$$

Yields a general technique: break integrals apart at x = 1.

Inclusions and relationships:

$$m(X) < \infty \implies L^{\infty} \subset L^2 \subset L^1$$

 $\ell^2(\mathbb{Z}) \subset \ell^1(\mathbb{Z}) \subset \ell^{\infty}(\mathbb{Z}).$

7.4.1 a

Theorem (Holder's Inequality):

$$||fg||_1 \le ||f||_p ||g||_q$$
.

Proof:

It suffices to show this when $\|f\|_p = \|g\|_q = 1,$ since

$$||fg||_1 \le ||f||_p ||f||_q \Longleftrightarrow \int \frac{|f|}{||f||_p} \frac{|g|}{||g||_q} \le 1.$$

Using $AB \leq \frac{1}{p}A^p + \frac{1}{q}B^q$, we have

$$\int |f||g| \le \int \frac{|f|^p}{p} \frac{|g|^q}{q} = \frac{1}{p} + \frac{1}{q} = 1 \blacksquare.$$

Theorem (Minkowski's Inequality):

$$||f + g||_p \le ||f||_p + ||g||_p.$$

Proof:

We first note

$$|f+g|^p = |f+g||f+g|^{p-1} \le (|f|+|g|)|f+g|^{p-1}.$$

Note that if p, q are conjugate exponents then

$$\frac{1}{q} = 1 - \frac{1}{p} = \frac{p-1}{p}$$
$$q = \frac{p}{p-1}.$$

Then taking integrals yields

$$\begin{split} \|f+g\|_p^p &= \int |f+g|^p \\ &\leq \int (|f|+|g|) |f+g|^{p-1} \\ &= \int |f||f+g|^{p-1} + \int |g||f+g|^{p-1} \\ &= \left\|f(f+g)^{p-1}\right\|_1 + \left\|g(f+g)^{p-1}\right\|_1 \\ &\leq \|f\|_p \left\|(f+g)^{p-1}\right\|_q + \|g\|_p \left\|(f+g)^{p-1}\right\|_q \\ &= \left(\|f\|_p + \|g\|_p\right) \left\|(f+g)^{p-1}\right\|_q \\ &= \left(\|f\|_p + \|g\|_p\right) \left(\int |f+g|^{(p-1)q}\right)^{\frac{1}{q}} \\ &= \left(\|f\|_p + \|g\|_p\right) \left(\int |f+g|^p\right)^{1-\frac{1}{p}} \\ &= \left(\|f\|_p + \|g\|_p\right) \frac{\int |f+g|^p}{\left(\int |f+g|^p\right)^{\frac{1}{p}}} \\ &= \left(\|f\|_p + \|g\|_p\right) \frac{\|f+g\|_p}{\|f+g\|_p} \end{split}$$

and canceling common terms yields

$$1 \le \left(\|f\|_p + \|g\|_p \right) \frac{1}{\|f + g\|_p}$$

$$\implies \|f + g\|_p \le \|f\|_p + \|g\|_p \blacksquare.$$

7.4.2 c

Definition (Infinity Norm):

$$\begin{split} L^{\infty}(X) &= \left\{ f: X \longrightarrow \mathbb{C} \ \middle| \ \|f\|_{\infty} < \infty \right\} \\ \text{where} \\ \|f\|_{\infty} &= \inf_{\alpha \geq 0} \left\{ \alpha \ \middle| \ m \left\{ |f| \geq \alpha \right\} = 0 \right\}. \end{split}$$

Theorem:

$$m(X) < \infty \implies \lim_{p \longrightarrow \infty} \|f\|_p = \|f\|_{\infty}.$$

Proof: Let $M = ||f||_{\infty}$. For any L < M, let $S = \{|f| \ge L\}$. Then m(S) > 0 and

$$||f||_{p} = \left(\int_{X} |f|^{p}\right)^{\frac{1}{p}}$$

$$\geq \left(\int_{S} |f|^{p}\right)^{\frac{1}{p}}$$

$$\geq L \ m(S)^{\frac{1}{p}} \xrightarrow{p \longrightarrow \infty} L$$

$$\implies \liminf_{p} ||f||_{p} \geq M.$$

We also have

$$||f||_p = \left(\int_X |f|^p\right)^{\frac{1}{p}}$$

$$\leq \left(\int_X M^p\right)^{\frac{1}{p}}$$

$$= M \ m(X)^{\frac{1}{p}} \xrightarrow{p \to \infty} M$$

$$\implies \limsup_p ||f||_p \leq M \blacksquare.$$

Note: this doesn't help with this problem at all.

Solution:

$$\int f^{p} = \int_{x \le 1} f^{p} + \int_{x=1} f^{p} + \int_{x \ge 1} f^{p}$$

$$= \int_{x \le 1} f^{p} + \int_{x=1} 1 + \int_{x \ge 1} f^{p}$$

$$= \int_{x \le 1} f^{p} + m(\{f = 1\}) + \int_{x \ge 1} f^{p}$$

$$\xrightarrow{p \to \infty} 0 + m(\{f = 1\}) + \begin{cases} 0 & m(\{x \ge 1\}) = 0\\ \infty & m(\{x \ge 1\}) > 0. \end{cases}$$

7.5 5: Dual Spaces

Definition: A map $L: X \longrightarrow \mathbb{C}$ is a linear functional iff

$$L(\alpha \mathbf{x} + \mathbf{y}) = \alpha L(\mathbf{x}) + L(\mathbf{y}).$$

Theorem (Riesz Representation for Hilbert Spaces): If Λ is a continuous linear functional on a Hilbert space H, then there exists a unique $y \in H$ such that

$$\forall x \in H, \quad \Lambda(x) = \langle x, y \rangle.$$

Proof:

- Define $M := \ker \Lambda$.
- \bullet Then M is a closed subspace and so $H=M\oplus M^\perp$
- There is some $z \in M^{\perp}$ such that ||z|| = 1.
- Set $u := \Lambda(x)z \Lambda(z)x$
- Check

$$\Lambda(u) = \Lambda(\Lambda(x)z - \Lambda(z)x) = \Lambda(x)\Lambda(z) - \Lambda(z)\Lambda(x) = 0 \implies u \in M$$

• Compute

$$\begin{split} 0 &= \langle u, \ z \rangle \\ &= \langle \Lambda(x)z - \Lambda(z)x, \ z \rangle \\ &= \langle \Lambda(x)z, \ z \rangle - \langle \Lambda(z)x, \ z \rangle \\ &= \Lambda(x)\langle z, \ z \rangle - \Lambda(z)\langle x, \ z \rangle \\ &= \Lambda(x)\|z\|^2 - \Lambda(z)\langle x, \ z \rangle \\ &= \Lambda(x) - \Lambda(z)\langle x, \ z \rangle \\ &= \Lambda(x) - \langle x, \ \overline{\Lambda(z)}z \rangle, \end{split}$$

- Choose $y := \overline{\Lambda(z)}z$.
- Check uniqueness:

$$\langle x, y \rangle = \langle x, y' \rangle \quad \forall x$$

$$\implies \langle x, y - y' \rangle = 0 \quad \forall x$$

$$\implies \langle y - y', y - y' \rangle = 0$$

$$\implies ||y - y'|| = 0$$

$$\implies y - y' = \mathbf{0} \implies y = y'.$$

7.5.1 b

Theorem (Continuous iff Bounded): Let $L: X \longrightarrow \mathbb{C}$ be a linear functional, then the following are equivalent:

- 1. L is continuous
- 2. L is continuous at zero
- 3. L is bounded, i.e. $\exists c \geq 0 \mid |L(x)| \leq c||x||$ for all $x \in H$
- $2 \implies 3$: Choose $\delta < 1$ such that

$$||x|| \le \delta \implies |L(x)| < 1.$$

Then

$$|L(x)| = \left| L\left(\frac{\|x\|}{\delta} \frac{\delta}{\|x\|} x\right) \right|$$
$$= \frac{\|x\|}{\delta} \left| L\left(\delta \frac{x}{\|x\|}\right) \right|$$
$$\leq \frac{\|x\|}{\delta} 1,$$

so we can take $c = \frac{1}{\delta}$.

 $3 \implies 1$:

We have $|L(x-y)| \le c||x-y||$, so given $\varepsilon \ge 0$ simply choose $\delta = \frac{\varepsilon}{c}$.

7.5.2 c

Definition (Dual Space):

$$X^{\vee} := \left\{ L : X \longrightarrow \mathbb{C} \mid L \text{ is continuous } \right\}$$

Definition (Operator Norm):

$$||L||_{X^{\vee}} \coloneqq \sup_{\substack{x \in X \\ ||x|| = 1}} |L(x)|$$

Theorem: (Operator Norm is a Norm)

Proof: The only nontrivial property is the triangle inequality, but

$$||L_1 + L_2|| = \sup |L_1(x) + L_2(x)| \le \sup L_1(x) + \sup L_2(x) = ||L_1|| + ||L_2||.$$

Theorem (Completeness in Operator Norm):

 X^{\vee} equipped with the operator norm is a Banach space.

Proof:

- Let $\{L_n\}$ be Cauchy in X^{\vee} .
- Then for all $x \in C$, $\{L_n(x)\}\subset \mathbb{C}$ is Cauchy and converges to something denoted L(x).
- Need to show L is continuous and $||L_n L|| \longrightarrow 0$.
- Since $\{L_n\}$ is Cauchy in X^{\vee} , choose N large enough so that

$$n, m \ge N \implies ||L_n - L_m|| < \varepsilon \implies |L_m(x) - L_n(x)| < \varepsilon \quad \forall x \mid ||x|| = 1.$$

• Take $n \longrightarrow \infty$ to obtain

$$m \ge N \implies |L_m(x) - L(x)| < \varepsilon \quad \forall x \mid ||x|| = 1$$

$$\implies ||L_m - L|| < \varepsilon \longrightarrow 0.$$

• Continuity:

$$|L(x)| = |L(x) - L_n(x) + L_n(x)|$$

$$\leq |L(x) - L_n(x)| + |L_n(x)|$$

$$\leq \varepsilon ||x|| + c||x||$$

$$= (\varepsilon + c)||x|| \blacksquare.$$

8 Exam 2 (2018)

Link to PDF File

9 Exam 2 (2014)

Link to PDF File

10 Qual: Fall 2019

10.1 1

See phone photo?

10.2 2

DCT?

10.3 3

"Follow your nose."

10.4 4

See Problem Set 8.

Bessel's Inequality: For any orthonormal set in a Hilbert space (not necessarily a basis), we have

$$\sum_{n=1}^{\infty} |\langle x, u_n \rangle|^2 \le ||x||^2$$

Proof:

$$0 \le \left\| x - \sum_{k=1}^{n} \left\langle x, e_k \right\rangle e_k \right\|^2$$

Corollary (Parseval's Identity): If span $\{u_n\}$ is dense in \mathcal{H} , so it is a basis, then this is an equality.

Riesz-Fischer: Let $U = \{u_n\}_{n=1}^{\infty}$ be an orthonormal set (not necessarily a basis), then

1. There is an isometric surjection

$$\mathcal{H} \longrightarrow \ell^2(\mathbb{N})$$

 $\mathbf{x} \mapsto \{\langle \mathbf{x}, \mathbf{u}_n \rangle\}_{n=1}^{\infty}$

i.e. if $\{a_n\} \in \ell^2(\mathbb{N})$, so $\sum |a_n|^2 < \infty$, then there exists a $\mathbf{x} \in \mathcal{H}$ such that

$$a_n = \langle \mathbf{x}, \mathbf{u}_n \rangle \quad \forall n.$$

2. \mathbf{x} can be chosen such that

$$\|\mathbf{x}\|^2 = \sum |a_n|^2$$

Note: the choice of **x** is unique \iff $\{u_n\}$ is **complete**, i.e. $\langle \mathbf{x}, \mathbf{u}_n \rangle = 0$ for all n implies

Proof:

• Given $\{a_n\}$, define $S_N = \sum_{n=1}^N a_n \mathbf{u}_n$. • S_N is Cauchy in \mathcal{H} and so $S_N \longrightarrow \mathbf{x}$ for some $\mathbf{x} \in \mathcal{H}$. • $\langle x, u_n \rangle = \langle x - S_N, u_n \rangle + \langle S_N, u_n \rangle \longrightarrow a_n$ • By construction, $\|x - S_N\|^2 = \|x\|^2 - \sum_{n=1}^N |a_n|^2 \longrightarrow 0$, so $\|x\|^2 = \sum_{n=1}^\infty |a_n|^2$.

10.5 5

See Problem Set 5.

Heine-Cantor theorem: Every continuous function on a compact set is uniformly continuous. Uniform continuity:

$$\forall \varepsilon \quad \exists \delta(\varepsilon) \mid \quad \forall x, y, \quad |x - y| < \delta \implies |f(x) - f(y)| < \varepsilon$$

$$\iff \forall \varepsilon \quad \exists \delta(\varepsilon) \mid \quad \forall x, y, \quad |y| < \delta \implies |f(x - y) - f(y)| < \varepsilon$$

Fubini-Tonelli interchange of integrals, where the change of bounds becomes very important.

Continuity in L^1 :

$$\lim_{y \longrightarrow 0} \left\| \tau_y f - f \right\|_1 = 0.$$

10.6 Definitions

- Banach Space
- L^p

10.7 Useful Results

- Cauchy-Schwarz
- Young's Inequality
- Holder's Inequality
- Minkowski's Inequality
- Jensen's Inequality:

$$r^{-1} \coloneqq p^{-1} + q^{-1} - 1 \implies \|f * g\|_r \le \|f\|_p \|g\|_q$$

- Useful variant take q=1 to get $\|f*g\|_p \leq \|f\|_p \|g\|_1$
- Take p=1 to show L_1 is closed under *.
- The Riemann-Lebesgue Lemma
- Proving that $\delta \notin L_1(\mathbb{R})$ and that there is no such identity
 - Rather, is a distribution or measure that acts on f and satisfies $f(x) \int_{\mathbb{R}} f(t) \delta(t-x) dt$
- Fubini's Theorem
- Density Results:
 - $C_c(\mathbb{R}) \subset C_0(\mathbb{R})$
- $C_c(\mathbb{R}) \cap C^{\infty}(\mathbb{R}) \neq \emptyset$, e.g. take $f(x) = e^{\frac{-1}{x^2}} \chi_{(0,\infty}(x)$.
- The Banach Algebra $L^1(\mathbb{R})$ is not a principal ideal domain.
- \bullet Every locally compact abelian group G has a unique Borel measure (up to scaling) that is positive, regular, translation-invariant (the Haar measure).
 - For \mathbb{R} , $(S_1)^2$, equal to the Lebesgue measure. For \mathbb{Z} , the counting measure.