# 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,

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

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

Note: doesn't work for  $\ell$  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:** An  $F_{\sigma}$  set is a union of closed sets, and a  $G_{\delta}$  set is an intersection of opens.

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

### Lemma 2.5.

Singleton sets in  $\mathbb{R}$  are closed, and thus  $\mathbb{Q}$  is an  $F_{\sigma}$  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  $F_\sigma$  set, which follows by considering the oscillation  $\omega_f$ .  $\omega_f(x) = 0 \iff f$  is continuous at x, and  $D_f = \bigcup_n A_{\frac{1}{n}}$  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 (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  $f(x) := \lim_{k \longrightarrow \infty} f_k(x) < \infty$  exists almost everywhere.

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

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

**Theorem (Important Example):** 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:* 

Step  $\theta$ : Let  $\{f_k\}$  be Cauchy in X.

Step 1: 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  $f(x) := \lim_k f_k(x)$ .

Step 2: Show that  $||f_k - f|| \longrightarrow 0$ :

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

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  $\varepsilon/2$ .

Step 3: 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 3.1.

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

Corollary: 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 3.2.

A uniform limit of continuous functions is continuous.

**Lemma (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 3.3(Baby Commuting Limits with Integrals).

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

## Lemma 3.4(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 3.5 (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 3.6.

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

**Lemma (p-tests):** Let n be a fixed dimension and set  $B = \{x \in \mathbb{R}^n \mid ||x|| \le 1\}$ .

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

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

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

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

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

3 UNIFORM CONVERGENCE

## 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  $E = \bigcup B_n(0) \cap E$ 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**: 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 ⊆ F ⇒ m<sub>\*</sub>(E) ≤ m<sub>\*</sub>(F).
  Countable Subadditivity: m<sub>\*</sub>(∪E<sub>i</sub>) ≤ ∑ m<sub>\*</sub>(E<sub>i</sub>).
  Approximation: For all E there exists a G ⊇ E such that m<sub>\*</sub>(G) ≤ m<sub>\*</sub>(E) + ε.
  Disjoint\* Additivity: m<sub>\*</sub>(A ∐B) = m<sub>\*</sub>(A) + m<sub>\*</sub>(B).

Note: this holds for outer measure **iff** dist(A, B) > 0.

### Lemma 4.3 (Subtraction of Measure):).

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

### Lemma 4.4(Continuity of Measure).

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

Proof.

- 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  $E_1 = (\coprod E_j \setminus E_{j+1}) \coprod (\bigcap E_j)$ , 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.

## Lemma (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:  $F_{\sigma}$  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  $N := E \setminus K$ , then m(N) = 0.

### Lemma 4.7.

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

## Lemma 4.8.

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

 $\it Proof:$  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

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

## Application:

$$m\left(\left\{x \text{ such that } \exists \text{ inf. many } \frac{p}{q} \text{ with } \left|x-\frac{p}{q}\right| \leq \frac{1}{q^3}\right\}\right) = 0.$$

## Proof.

Idea: write  $E_j$  to be the above set with p,q replaced by  $p_j,q_j$  where  $r_j = \frac{p_j}{a_i}$  is an enumeration

of  $\mathbb{Q}$ , then  $m(E_j) \leq \frac{2}{q^3}$  and  $\sum \frac{1}{q^3} < \infty$ .

- If  $E = \limsup E_j$  with  $\sum m(E_j) < \infty$  then m(E) = 0.
- If  $E_j$  are measurable, then  $\limsup_{j \to j} E_j$  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.  $E = \limsup_{j} E_{j} = \bigcap_{k=1}^{\infty} \bigcup_{j=k}^{\infty} E_{j} \implies E \subseteq \bigcup_{j=k}^{\infty} \text{ for all } k$   $E \subset \bigcup_{j=k}^{\infty} \implies m(E) \le \sum_{j=k}^{\infty} m(E_{j}) \stackrel{k \longrightarrow \infty}{\longrightarrow} 0$ .

## 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].

## Integration

**Definition:**  $f \in L^+$  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:** A measurable function is integrable iff  $||f||_1 < \infty$ .

Useful facts about  $C_c$  functions:

- Bounded almost everywhere
- Uniformly continuous

## 5.1 Convergence Theorems

Monotone Convergence Theorem (MCT):

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

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

Needs to be positive and increasing.

Dominated Convergence Theorem (DCT):

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

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

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:** 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 (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 (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 (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.

Differentiating under the integral:

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

$$\frac{\partial}{\partial t} F(t) := \lim_{h \to 0} \int \frac{f(x, t+h) - f(x, t)}{h} dx$$

$$\stackrel{\text{DCT}}{=} \int \frac{\partial}{\partial t} f(x, t) dx.$$

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  $f_k(\xi,t) \in L_1$ .

**Lemma (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:** 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 
$$F_N = \sum_{k=0}^{N} f_k$$
 and  $F = \lim_{k \to \infty} F_k$ , then  $||F_N||_1 \le \sum_{k=0}^{N} ||f_k|| < \infty$  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 (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.

## Lemma (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 ( $L^1$  functions may 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 ( $L^1$ functions have Small Tails):

If  $f \in L^1$ , then for every  $\varepsilon$  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 ( $L^1$ 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 ( $L^1$ functions are finite almost everywhere):

If  $f \in L^1$ , then  $m(\{f(x) = \infty\}) = 0$ .

Proof (Split up domain2): 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  $g \in C_c$ .

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

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

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

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  $C_c^{\infty}$

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

- 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

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

**⇒** ·

- 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

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

### 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.  $f, g \in L^1, f \leq M, g \leq M' \implies f * g \xrightarrow{x \to \infty} 0.2$

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

### 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:* 

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

## 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:

- $f \in L^1 \implies \hat{f}$  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):

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

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

Use the following facts:

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

- $\widehat{g}_t(x) = c$  g(x/c) c e•  $\widehat{g}_t(x) = g(tx) = e^{-\pi t^2 |x|^2}$   $\phi(\xi) \coloneqq e^{2\pi i x \cdot \xi} \ \widehat{g}_t(\xi)$   $\widehat{\phi}(\xi) = \mathcal{F}(\widehat{g}_t(\xi x)) = g_t(x \xi)$   $\lim_{t \to 0} \phi(\xi) = e^{2\pi i x \cdot \xi}$

Take the modified integral:

$$\begin{split} I_t(x) &= \int \widehat{f}(\xi) \ e^{2\pi i x \cdot \xi} \ e^{-\pi t^2 |\xi|^2} \\ &= \int \widehat{f}(\xi) \phi(\xi) \\ &= \int f(\xi) \widehat{\phi}(\xi) \\ &= \int f(\xi) \widehat{\widehat{g}}(\xi - x) \\ &= \int f(\xi) g_t(x - \xi) \ d\xi \\ &= \int f(y - x) g_t(y) \ dy \quad (\xi = y - x) \\ &= (f * g_t) \\ &\longrightarrow f \text{ in } L^1 \text{ as } t \longrightarrow 0, \end{split}$$

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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:

$$\begin{split} \int f^p &= \int_{x \le 1} f^p + \int_{x=1} f^p + \int_{x \ge 1} f^p \\ &= \int_{x \le 1} f^p + \int_{x=1} 1 + \int_{x \ge 1} f^p \\ &= \int_{x \le 1} f^p + m(\{f=1\}) + \int_{x \ge 1} f^p \\ &\xrightarrow{p \longrightarrow \infty} 0 + m(\{f=1\}) + \begin{cases} 0 & m(\{x \ge 1\}) = 0 \\ \infty & m(\{x \ge 1\}) > 0. \end{cases} \end{split}$$

## 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  $\Lambda$  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<sup>p</sup>

### 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.