# Functional analysis

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# Part 1

**PRELIMINARIES** 

### 1.1 Relationships between structures

Let *X* be a set.

#### **Definition 1.1**

- 1. A basis of a topology is is a collection  $\mathcal B$  of subsets of X satisfying the following properties:
  - *i.* (cover) The base elements cover X.
  - ii. (intersection) For every  $B_1, B_2 \in \mathcal{B}$ , if  $x \in B_1 \cap B_2$ , then there is a  $B_x \in \mathcal{B}$  such that  $x \in B_x \subseteq B_1 \cap B_2$ .
- 2. A metric is a function  $d(\cdot, \cdot): X \times X \to [0, \infty)$  such that for all vectors  $x, y, z \in X$ , we have
  - *i.* (identity of indiscernibles) d(x, y) = 0 iff x = y.
  - ii. (symmetry) d(x,y) = d(y,x).
  - iii. (triangle inequality)  $d(x,z) \le d(x,y) + d(y,z)$ .
- 3. A norm is a function  $\|\cdot\|: X \to [0, \infty)$  such that for all vectors  $x, y \in X$  and scalar  $\alpha \in \mathbb{C}$ , we have
  - *i.* (identity of indiscernibles) ||x|| = 0 iff x = 0.
  - ii. (scaling)  $\|\alpha x\| = |\alpha| \|x\|$ .
  - iii. (triangle inequality)  $||x + y|| \le ||x|| + ||y||$ .
- 4. An inner product is a function  $\langle \cdot, \cdot \rangle : X \times X \to \mathbb{C}$  such that for all vectors  $x, y, z \in X$  and scalar  $\alpha \in \mathbb{C}$ , and we have
  - *i.* (positive-definiteness)  $\langle x, x \rangle \ge 0$  and  $\langle x, x \rangle = 0$  iff x = 0.
  - ii. (conjugate symmetry a.k.a. Hermitian)  $\langle x, y \rangle = \overline{\langle y, x \rangle}$ .
  - *iii.* (sesquilinearity)  $\langle \alpha x + y, z \rangle = \alpha \langle x, z \rangle + \langle y, z \rangle$ .

**Proposition 1.2** *Inner product*  $\Longrightarrow$  *norm*  $\Longrightarrow$  *metric*  $\Longrightarrow$  *topology.* 

#### Proof.

- A. *inner product*  $\Longrightarrow$  *norm*. Define the norm as  $\|\cdot\| = \sqrt{\langle \cdot, \cdot \rangle}$ .
  - i.  $||x|| = 0 \iff ||x||^2 = 0 \iff \langle x, x \rangle = 0 \iff x = 0 \text{ using 4.i.}$
  - ii.  $\|\alpha x\|^2 = \langle \alpha x, \alpha x \rangle = \langle \overline{\alpha} \alpha x, x \rangle = \overline{\alpha} \alpha \langle x, x \rangle = |\alpha|^2 \|x\|^2$  using 4.ii and 4.iii.
  - iii.

$$||x + y||^{2} = \langle x + y, x + y \rangle$$

$$= \langle x, x \rangle + \langle x, y \rangle + \langle y, x \rangle + \langle y, y \rangle \qquad [4.iii]$$

$$= ||x||^{2} + \langle x, y \rangle + \overline{\langle x, y \rangle} + ||y||^{2} \qquad [4.ii]$$

$$= ||x||^{2} + 2\Re \langle x, y \rangle + ||y||^{2}$$

$$\leq ||x||^{2} + 2|\langle x, y \rangle| + ||y||^{2}$$

$$\leq ||x||^{2} + 2||x|| ||y|| + ||y||^{2} \qquad [1.2]$$

$$= (||x|| + ||y||)^{2}.$$

- B. *norm*  $\Longrightarrow$  *metric*. Define the metric as d(x,y) = ||x-y||.
  - i.  $d(x, y) = 0 \iff ||x y|| = 0 \iff x y = 0 \iff x = y \text{ using 3.i.}$
  - ii. d(x,y) = ||x y|| = ||-(y x)|| = |-1| ||y x|| = d(y,x) using 3.ii.
  - iii.  $d(x,z) = ||x-z|| = ||(x-y) + (y-z)|| \le ||x-y|| + ||y-z|| = d(x,y) + d(y,z)$  using 3.iii.
- C.  $metric \implies topology$ . Define the basis of the topology as open balls of the form

$$D_r(x_0) = \{x \in X \mid d(x, x_0) < r\}, \quad x_0 \in X, r > 0.$$

That is,  $\mathcal{T} = \tau(\mathcal{B})$ , where  $\mathcal{B} = \{D_r(x_0) \mid x_0 \in X, r > 0\}$ .

All we have to do is show that  $\mathcal{B}$  is a basis. The cover is obvious. Note that for any  $B_1, B_2 \in \mathcal{B}$ , we can write  $B_1 = D_{r_1}(x_1), B_2 = D_{r_2}(x_2)$ . Suppose  $x \in B_1 \cap B_2$ . Then  $x \in D_r(x) \subseteq D_{r_1}(x_1) \cap D_{r_2}(x_2)$  if  $r \le \min\{r_1 - d(x, x_1), r_2 - d(x, x_2)\}$ , and we are done.

The topology induced by the metric is called the *metric topology*.

## 1.2 Strong, weak and weak\* convergence

Disclaimer: This section is shamelessly copied from Christopher Heil's notes.

**Definition 2.1** Let X be a normed vector space, and  $x_n, x \in X$ . We define the following convergences as  $n \to \infty$ .

$$(strong) x_n \to x \Longleftrightarrow ||x_n - x|| \to 0$$

$$(weak) x_n \overset{w}{\to} x \Longleftrightarrow \forall \phi \in X^*, (x_n - x, \phi) \to 0$$

**Definition 2.2** Let X be a normed vector space, and  $\phi_n, \phi \in X^*$ . We define the following convergences as  $n \to \infty$ .

$$\begin{array}{lll} (strong) & \phi_n \to \phi & \iff & \|\phi_n - \phi\| \to 0 \\ \\ (weak) & \phi_n \overset{w}{\to} \phi & \iff & \forall \xi \in X^{**}, \quad (\phi_n - \phi, \xi) \to 0 \\ \\ (weak^*) & \phi_n \overset{w^*}{\to} \phi & \iff & \forall x \in X, \quad (x, \phi_n - \phi) \to 0 \end{array}$$

**Remark 2.3** *Weak\* convergence is simply* pointwise convergence *for the functionals*  $\phi_n$ .

**Proposition 2.4** (strong  $\Rightarrow$  weak\* for convergence) Suppose  $\phi_n$ ,  $\phi \in X^*$ .

Then 
$$\phi_n \to \phi \Longrightarrow \phi_n \stackrel{w}{\to} \phi \Longrightarrow \phi_n \stackrel{w^*}{\to} \phi$$
.

*The second implication reverses if X is reflexive.* 

*Proof.* strong 
$$\Longrightarrow$$
 weak:  $(x_n - x, \phi) \le ||x_n - x|| ||\phi|| \to 0.$  weak  $\Longrightarrow$  weak\*:  $(x, \phi_n - \phi) = (\phi_n - \phi, x^{**}) \to 0.$ 

The claim about the reverse implication is now obvious.

Counterexample for converse of the first implication: Suppose  $X = \ell^2(\mathbb{N})$ . Then  $e_n \stackrel{w}{\to} 0$ , but  $||e_n - 0|| = 1 \to 0$ .

**Proposition 2.5** In Hilbert spaces, weak convergence plus convergence of norms  $(\|x_n\| \to \|x\|)$  is equivalent to strong convergence.

Proof. 
$$||x_n - x||^2 = \langle x_n - x, x_n - x \rangle = \langle x_n - x, x_n \rangle - \langle x_n - x, x \rangle \to 0.$$

**Proposition 2.6** Let H and K be Hilbert spaces, and let  $T \in B(H, K)$  be a compact operator.

Show that 
$$x_n \stackrel{w}{\to} x \Longrightarrow Tx_n \to Tx$$
.

Thus, a compact operator maps weakly convergent sequences to strongly convergent sequences.

Proof. Disclaimer: Stolen from MSx1142451.

 $Tx_n \stackrel{w}{\to} Tx$  by continuity. Thus if any subsequence has a strong limit, it certainly is Tx. But compactness guarantees every subsequence has a subsequence that converges to something: that something is Tx by uniqueness, and so by our above equivalence with convergence, we have  $Tx_n \to Tx$ .

# Part 2

HILBERT SPACES

### 2.1 Basics

In what follows,  $(H, \langle \cdot, \cdot \rangle)$  is a Hilbert space, and we write  $x \perp y$  iff  $\langle x, y \rangle = 0$ .

**Theorem 1.1** (Pythagorean) If  $x, y \in H$  and  $x \perp y$ , then  $||x + y||^2 = ||x||^2 + ||y||^2$ .

Proof.

$$||x + y||^2 = \langle x + y, x + y \rangle$$

$$= \langle x, x \rangle + \langle x, y \rangle + \langle y, x \rangle + \langle y, y \rangle$$

$$= ||x||^2 + ||y||^2.$$

**Theorem 1.2** (Cauchy–Schwarz inequality) If  $x, y \in H$ , then  $|\langle x, y \rangle| \le ||x|| ||y||$ .

*Proof.* (norm expansion) Note that  $0 \le \|x - \lambda y\|^2 = \|x\|^2 - 2\Re\left(\overline{\lambda}\langle x, y\rangle\right) + |\lambda|^2 \|y\|^2$ , so if we take  $\lambda = \frac{\langle x, y\rangle}{\|y\|^2}$ , we get  $0 \le \|x\|^2 - \frac{|\langle x, y\rangle|^2}{\|y\|^2}$ , which gives us the required result.  $\square$ 

*Proof.* (*projection*) Note that we can write  $x = x_{\parallel} + x_{\perp}$ , where  $x_{\parallel}$  is the component of x in the direction of y and  $x_{\perp}$  is the component of x in the direction perpendicular to y. Explicitly,  $x_{\parallel} = \langle x, \hat{y} \rangle \hat{y} = \langle x, y \rangle \frac{y}{\|y\|^2}$ . Using the Pythagorean theorem (1.1), we get

$$||x||^2 = ||x_{\parallel}||^2 + ||x_{\perp}||^2 \ge ||x_{\parallel}||^2 = \frac{|\langle x, y \rangle|^2}{||y||^2}.$$

**Theorem 1.3 (Riesz–Fischer)**  $L^p(X, \mu)$  *is complete for*  $p \in [0, \infty]$ .

*Proof.* Let  $(f_n)$  be a Cauchy sequence in  $L^p$ . We have to show that there exists  $f \in L^p$  such that  $f_n \to f$  in  $L^p$ .

Since  $(f_n)$  is Cauchy, for every  $\varepsilon > 0$ , there exists  $N \in \mathbb{N}$  such that for every n, m > N, we have  $\|f_n - f_m\|_p < \varepsilon$ . Therefore, there exists a subsequence  $(f_{n_k})$  such that  $\|f_{n_{k+1}} - f_{n_k}\|_p < 2^{-(k+1)}$  for every  $k \in \mathbb{N}_0$ , where we adopt the convention that  $f_{n_0} \equiv 0$ .

Note that  $f_{n_k} = \sum_{i=0}^{k-1} (f_{n_{i+1}} - f_{n_i})$  for each  $k \in \mathbb{N}$ .

Define  $f = \sum_{j=0}^{\infty} \left( f_{n_{j+1}} - f_{n_j} \right)$ . Clearly,  $f_{n_k} \to f$  pointwise. Moreover, if  $g = \sum_{j=0}^{\infty} \left| f_{n_{j+1}} - f_{n_j} \right|$ , then  $\left| f_{n_k} \right| \le g$  and  $\left\| g \right\|_p \le \sum_{j=0}^{\infty} \left\| f_{n_{j+1}} - f_{n_j} \right\|_p \le 1$  using the triangle inequality. Therefore, by Lebesgue's dominated convergence theorem,  $f_{n_k} \to f$  in  $L^p$ . Similar to g, we get  $\left\| f \right\|_p \le 1$ , showing  $f \in L^p$ . All that is left to show is that  $f_n \to f$  in  $L^p$ . Using the fact that the sequence is Cauchy, we get

$$||f_n - f||_p \le ||f_n - f_{n_k}||_p + ||f_{n_k} - f||_p \to 0 \text{ as } n \to \infty.$$

# Part 3

OPERATOR THEORY

### 3.1 Elementary ideas

A great source is Trace class operators and Hilbert-Schmidt operators by Jordan Bell.

#### Intuition 3.1.1

On a separable Hilbert space, we have

- $ightharpoonup T \in \mathcal{B}^{\infty} \Longleftrightarrow \lambda \in \ell^{\infty} \text{ (bounded)}$ Example  $I: \ell^2 \to \ell^2: e_n \mapsto e_n$ .
- $ightharpoonup T \in \mathcal{B}_0 \Longleftrightarrow \lambda \in c_0 \text{ (compact)}$
- Example  $T: \ell^2 \to \ell^2: e_n \mapsto \frac{1}{\sqrt{n}} e_n$ .
- $ightharpoonup T \in \mathbb{B}^2 \Longleftrightarrow \lambda \in \ell^2 \text{ (Hilbert-Schmidt)}$ Example  $T: \ell^2 \to \ell^2: e_n \mapsto \frac{1}{n}e_n$ .
- $ightharpoonup T \in \mathcal{B}^1 \Longleftrightarrow \lambda \in \ell^1 \text{ (trace-class)}$ Example  $T: \ell^2 \to \ell^2: e_n \mapsto \frac{1}{n^2}e_n$ .
- $\begin{array}{c} \vdash T \in \mathcal{B}_{00} \Longleftrightarrow \lambda \in c_{00} \text{ (degenerate or finite rank)} \\ \text{Example } T: \ell^2 \to \ell^2: e_n \mapsto \alpha_n e_n \mathbb{1}_{[N]}(n) \text{ for } \alpha_n \in \mathbb{C} \text{ and } N \in \mathbb{N}. \end{array}$

Since the dual of  $c_0$  is  $\ell^1$  and the dual of  $\ell^1$  is  $\ell^{\infty}$ , we have  $\mathcal{B}_0^* = \mathcal{B}^1$  and  $(\mathcal{B}^1)^* = \mathcal{B}^{\infty}$ . Similarly,  $(\mathcal{B}^2)^* = \mathcal{B}^2$ .

Theorem 1.2 (Operator inclusions)  $\mathcal{B}_{00} \subset \mathcal{B}^1 \subset \mathcal{B}^2 \subset \mathcal{B}_0 \subset \mathcal{B}^{\infty}$ 

Proof.

- Trivial

- ((<BMC2009>), Proposition 4.6) If *T* is unbounded, we can find a sequence of unit vectors  $(e_n)$  such that  $||Te_n|| \nearrow \infty$ . So  $Te_n$  cannot have a convergent subsequence, for if  $Te_n \to x$ , then  $||Te_n|| \to ||x||$ .

**Proposition 1.3** For  $T \in \mathcal{B}^{\infty}$ ,  $||T||_{\infty} = \sup\{|\langle Tx, y \rangle|\} : ||x|| = 1$ , ||y|| = 1.

Proof.

(
$$\leq$$
) Since  $||Tx|| = \frac{||Tx||^2}{||Tx||} = \frac{\langle Tx, Tx \rangle}{||Tx||} = \langle Tx, \frac{Tx}{||Tx||} \rangle$ , we have  $||T||_{\infty} = \sup \{||Tx|| : ||x|| = 1\} \le \sup \{|\langle Tx, y \rangle| : ||x|| = 1, ||y|| = 1\}$ .

(
$$\geq$$
) Since  $\langle Tx, y \rangle \leq ||Tx|| ||y|| \leq ||T||_{\infty} ||x|| ||y||$ , we have 
$$\sup \{|\langle Tx, y \rangle| : ||x|| = 1, ||y|| = 1\} \leq ||T||_{\infty}.$$

# 3.1.2 Projection operators

**Proposition 1.4**  $||P||_{\infty} \leq 1$ .

*Proof.* Since 
$$||Px||^2 = \langle Px, Px \rangle = \langle P^*Px, x \rangle = \langle PPx, x \rangle = \langle Px, x \rangle \leq ||Px|| \, ||x||$$
, we have  $||P||_{\infty} \leq 1$ .

**Proposition 1.5** A projection operator is compact iff its image is finite dimensional.

#### Proof.

- (⇒) Let  $P: H \to H$  be a projection operator, so that  $P^2 = P$ , or P(P I) = 0.
- ( $\Leftarrow$ ) Since the image is finite dimensional, fix an orthonormal basis  $e_1, ..., e_n$  of im T.

### 3.2 Optimization

# 3.2.1 Duality in optimization is the same as duality in functional analysis

For an various intuitions of duality in optimization, see MSx223235.

Let X and Y be Banach spaces, and  $X^*$  and  $Y^*$  be their (algebraic?) duals. Consider the two problems, with  $\phi_0, y_0$  fixed. Here  $(\cdot, \cdot)$  denotes the canonical duality pairing.

See the following diagram for more details.

$$x \longmapsto \begin{array}{c} x \longmapsto T \\ x \in X & \xrightarrow{T} & y_0 \\ \downarrow & \downarrow \\ \phi_0, T^* \psi \in X^* & \xrightarrow{T^*} & y_0^* \\ & & \uparrow \\ T^* \psi & \xrightarrow{T^*} & \psi \end{array} \Rightarrow \psi$$

# **BIBLIOGRAPHY**