# INTRODUCTION TO ANALYSIS

# CARSON JAMES

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# Preface

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#### 1. Real and Complex Numbers

Note 1.0.1. As a starting point, we will take as fact the existence of the natural numbers

$$\mathbb{N} = \{1, 2, \cdots\}$$

the integers

$$\mathbb{Z} = \{\cdots, -2, -2, 0, 1, 2, \cdots\}$$

and the rational numbers

$$\mathbb{Q} = \left\{ \frac{a}{b} : a \in \mathbb{Z}, b \in \mathbb{N} \right\}$$

## 1.1. Real Numbers.

**Definition 1.1.1.** Let X be a set and  $\leq$  a relation on X. Then  $\leq$  is said to be a total **order** if for each  $a, b, c \in X$ ,

- $(1) \ a < a$
- (2)  $a \le b$  and  $b \le c$  implies that  $a \le c$
- (3)  $a \le b$  and  $b \le a$  implies that a = b
- (4)  $a \le b$  or  $b \le a$

**Exercise 1.1.2.** We define the relation  $\leq$  on  $\mathbb{Q}$  defined by

$$\frac{a}{b} \le \frac{c}{d}$$
 iff  $ad \le bc$ 

Then  $\leq$  is a total order of  $\mathbb{Q}$ .

*Proof.* Let  $\frac{a}{b}, \frac{c}{d}, \frac{e}{f} \in \mathbb{Q}$ . Then

- (1)  $\frac{a}{b} \leq \frac{a}{b}$  since  $ab \leq ab$ . (2) if  $\frac{a}{b} \leq \frac{c}{d}$  and  $\frac{c}{d} \leq \frac{e}{f}$ , then  $ad \leq bc$  and  $cf \leq de$ . Multiplying the first inequality by fand the second inequality by b, we obtain  $adf \leq bcf \leq bde$ . Dividing both sides by d yields  $af \leq be$ . Hence  $\frac{a}{b} \leq \frac{e}{f}$ .
- (3) if  $\frac{a}{b} \leq \frac{c}{d}$  and  $\frac{c}{d} \leq \frac{a}{b}$ , then  $ad \leq bc$  and  $bc \leq ab$ . This implies that ad = bc. Hence  $\frac{a}{b} = \frac{c}{d}$ .

#### 2. Metric Spaces

#### 2.1. Introduction.

**Definition 2.1.1.** Let M be a set and  $d: M \times M \to \mathbb{R}$ . Then d is said to be a **metric on** M if for each  $x, y, z \in M$ ,

- (1) d(x,y) = 0 iff x = y
- (2)  $d(x,y) \le d(x,z) + d(z,y)$

**Exercise 2.1.2.** Let M be a set and  $d: M \times M \to \mathbb{R}$  a metric on M. Then for each  $x, y \in M$ ,  $d(x, y) \ge 0$ .

*Proof.* Let  $x, y, z \in M$ . Then  $d(x, z) \leq d(x, y) + d(y, z)$ . This implies that  $d(x, z) - d(x, y) \leq d(y, z)$ . Since z is arbitrary, taking z = x, we obtain

$$d(x,x) - d(x,y) \le d(y,x) \implies -d(x,y) \le d(x,y)$$
$$\implies 0 \le 2d(x,y)$$
$$\implies d(x,y) \ge 0$$

**Definition 2.1.3.** Let M be a set and  $d: M \times M \to [0, \infty)$  a metric. Then (M, d) is called a **metric space**.

**Definition 2.1.4.** Let (M, d) be a metric space and  $A, B \subset M$ . We define the **distance** between A and B, denoted d(A, B), by

$$d(A,B) = \inf_{\substack{a \in A \\ b \in B}} d(a,b)$$

**Exercise 2.1.5.** Let (M,d) be a metric space. Then for each  $A,B\subset M$  and  $c\in M$ ,

$$d(A, B) < d(A, c) + d(c, B)$$

*Proof.* Let  $A, B \subset M$ ,  $c \in M$  and  $\epsilon > 0$ . Choose  $a \in A$  and  $b \in B$  such that  $d(a, c) < d(A, c) + \epsilon/2$  and  $d(c, b) < d(c, B) + \epsilon/2$ . Then

$$d(A, B) \le d(a, b)$$

$$\le d(a, c) + d(c, b)$$

$$< d(A, c) + \frac{\epsilon}{2} + d(c, B) + \frac{\epsilon}{2}$$

$$= d(A, c) + d(c, B) + \epsilon$$

Since  $\epsilon > 0$  is arbitrary,  $d(A, B) \leq d(A, c) + d(c, B)$ .

**Definition 2.1.6.** Let M be a set,  $d_1, d_2 : M \times M \to [0, \infty)$  metrics on M and p a property. Then  $d_1$  and  $d_2$  are said to be p-equivalent if  $\mathrm{id}_M : (M, d_1) \to (M, d_2)$  and  $\mathrm{id}_M : (M, d_2) \to (M, d_1)$  have property p. We say that  $d_1$  and  $d_2$  are **topologically equivalent** when p is the property of being continuous. We say that  $d_1$  and  $d_2$  are **equivalent** when p is the property of being Lipschitz.

**Definition 2.1.7.** Let  $(X, d_X)$  and  $(Y, d_Y)$  be metric spaces and  $f: X \to Y$ . Then f is said to be **Lipchitz** if there exists  $K \ge 0$  such that for each  $a, b \in X$ ,

$$d_Y(f(a), f(b)) \le K d_X(a, b)$$

**Exercise 2.1.8.** Let  $(X, d_X)$  and  $(Y, d_Y)$  be metric spaces and  $f: X \to Y$ . If f is Lipchitz, then f is uniformly continuous.

*Proof.* By definition, there exists  $K \geq 0$  such that for each  $a, b \in X$ ,

$$d_Y(f(a), f(b)) \le K d_X(a, b)$$

Let  $\epsilon > 0$ . Choose  $\delta = \epsilon/(K+1)$ . Let  $a, b \in X$ . Suppose that  $d_X(a, b) < \delta$ . Then

$$d_Y(f(a), f(b)) \le K d_X(a, b)$$

$$< K\delta$$

$$= K \frac{\epsilon}{K+1}$$

$$< \epsilon$$

**Definition 2.1.9.** Let  $(X, d_X)$  and  $(Y, d_Y)$  be metric spaces and  $f: X \to Y$  and  $x_0 \in X$ . Then f is said to be **locally Lipchitz at**  $x_0$  if there exists  $U \in \mathcal{N}_{x_0}$  such that f is Lipschitz on U.

**Definition 2.1.10.** Let  $(X, d_X)$  and  $(Y, d_Y)$  be metric spaces and  $f: X \to Y$ . Then f is said to be **locally Lipschitz** if for each  $x_0 \in X$ , f is locally Lipschitz at  $x_0$ .

**Definition 2.1.11.** Let X, Y be metric spaces and  $T: X \to Y$ . Then T is said to be an **isometry** if for each  $x_1, x_2 \in X$ ,  $d(Tx_1, Tx_2) = d(x_1, x_2)$ .

**Exercise 2.1.12.** Let X, Y be metric spaces and  $T: X \to Y$  and isometry. Then T is injective.

*Proof.* Let  $x_1, x_2 \in X$ . Suppose that  $Tx_1 = Tx_2$ . Then  $0 = d(Tx_1, Tx_2) = d(x_1, x_2)$ . So  $x_1 = x_2$ . Hence T is injective.  $\square$ 

**Note 2.1.13.** Let X, Y be metric spaces and  $T: X \to Y$  an isometry. Then T is clearly continuous. If T is surjective, then  $T^{-1}$  is an isometry and therefore continuous. Hence T is a homeomorphism.

**Definition 2.1.14.** Let (M, d) be a metric space. Then (M, d) is said to be a **Polish space** if (M, d) is complete and separable.

**Exercise 2.1.15.** Let (X, d) be a compact metric space,  $E \subset X$  closed,  $U \subset X$  open. Suppose that  $E \subset U$ . Then there exists  $\delta > 0$  such that for each  $x \in E$ ,  $B(x, \delta) \subset U$ .

Proof. Since X is compact, E and  $U^c$  are compact. Then there exist  $x_0 \in E$  and  $y_0 \in U^c$  such that  $d(E, U^c) = d(x_0, y_0)$ . Since  $E \cap U^c = \emptyset$ ,  $x_0 \neq y_0$  and  $d(E, U^c) > 0$ . Put  $\epsilon = d(E, U^c)$  and  $\delta = \frac{\epsilon}{2}$ . Let  $x \in E$ ,  $w \in B(x, \delta)$  and  $y \in U^c$ . Then

$$d(y, w) \ge d(y, x) - d(x, w)$$

$$> \epsilon - \delta$$

$$= \epsilon - \frac{\epsilon}{2}$$

$$= \frac{\epsilon}{2}$$

$$> 0$$

So  $y \neq w$ . Since and  $y \in U^c$  and  $w \in B(x, \delta)$  are arbitrary,  $B(x, \delta) \subset U$ .

**Definition 2.1.16.** Let X be a set, (M, d) a metrix space and  $B(S, M) = \{f : S \to M : f \text{ is bounded}\}$ . We define the **supremum metric**, denoted  $d_u : B(S, M) \times B(S, M) \to [0, \infty)$ , by

$$d_u(f,g) = \sup_{x \in X} d(f(x), g(x))$$

**Definition 2.1.17.** Let (X, d) be a metric space. Define

- (1)  $\operatorname{Aut}(X) = \{ \sigma : X \to X : \sigma \text{ is a homeomorphism} \}$
- (2)  $\operatorname{Aut}(X, d) = \{ \sigma : X \to X : \sigma \text{ is an isometric isomorphism} \}$

**Exercise 2.1.18.** Let (X,d) be a compact metric space,  $E \subset X$  closed,  $U \subset X$  open. Suppose that  $E \subset U$ . Let  $(f_n)_{n \in \mathbb{N}} \in \operatorname{Aut}(X)$ ,  $f \in \operatorname{Aut}(X)$ . Suppose that  $f_n \stackrel{\mathrm{u}}{\to} f$ . Then there exists  $N \in \mathbb{N}$  such that for each  $n \geq N$ ,  $f(E) \subset f_n(U)$ .

Proof. Since f is a homeomorphism, E is closed and U is open, f(E) is compact and f(U) is open and  $f(E) \subset f(U)$ . Then  $d(f(E), f(U^c)) > 0$ . Put  $\epsilon = d(f(E), f(U^c))$ . Choose  $\delta = \epsilon/2$ . Then there exists  $N \in \mathbb{N}$  such that for each  $n \in \mathbb{N}$ ,  $n \geq N$  implies that  $\sup_{z \in X} d(f(z), f_n(z)) < \delta$ .

Let  $n \geq N$ ,  $x \in E$  and  $w \in B(f(x), \delta)$ . For the sake of contradiction, suppose that  $w \in f_n(U^c)$ . Then there exist  $p \in U^c$  such that  $w = f_n(p)$ . Put  $z = f(p) \in f(U^c)$ . Then

$$\epsilon \le d(f(x), z)$$

$$\le d(f(x), w) + d(w, z)$$

$$= d(f(x), w) + d(f_n(p), f(p))$$

$$< \delta + \delta$$

$$= \epsilon$$

which is a contradiction. So  $w \in f_n(U)$ . Hence  $B(f(x), \delta) \subset f_n(U)$ 

# 2.2. Product Spaces.

#### 3. Topology

#### 3.1. Introduction.

**Definition 3.1.1.** Let X be a set and  $\mathcal{T} \subset \mathcal{P}(X)$ . Then  $\mathcal{T}$  is said to be a **topology on** X if

- $(1) X, \varnothing \in \mathcal{T}$
- (2) for each  $(U_{\alpha})_{\alpha \in A} \subset \mathcal{T}$ ,

$$\bigcup_{\alpha \in A} U_{\alpha} \in \mathcal{T}$$

(3) for each  $(U_j)_{j=1}^n \subset \mathcal{T}$ ,

$$\bigcap_{j=1}^{n} U_j \in \mathcal{T}$$

**Exercise 3.1.2.** Let X be a set and  $(\mathcal{T}_i)_{i\in I}$  a collection of topologies on X. Then  $\bigcap_{i\in I} \mathcal{T}_i$  is a topology on X.

Proof.

- (1) Since for each  $i \in I$ ,  $X, \emptyset \in \mathcal{T}_i$ , we have that  $X, \emptyset \in \bigcap_{i \in I} \mathcal{T}_i$ .
- (2) Let  $(U_{\alpha})_{\alpha \in A} \subset \bigcap_{i \in I} \mathcal{T}_i$ . Then for each  $i \in I$ ,  $(U_{\alpha})_{\alpha \in A} \subset T_i$ . So for each  $i \in I$ ,  $\bigcup_{\alpha \in A} U_{\alpha} \in \mathcal{T}_i$ . Thus  $\bigcup_{\alpha \in A} U_{\alpha} \in \bigcap_{i \in I} \mathcal{T}_i$ .
- (3) Let  $(U_j)_{j=1}^n \subset \bigcap_{i \in I} \mathcal{T}_i$ . Then for each  $i \in I$ ,  $(U_j)_{j=1}^n \subset T_i$ . So for each  $i \in I$ ,  $\bigcap_{j=1}^n U_j \in \mathcal{T}_i$ . Thus  $\bigcap_{j=1}^n U_j \in \bigcap_{i \in I} \mathcal{T}_i$ .

So  $\bigcap_{i \in I} \mathcal{T}_i$  is a topology on X.

**Definition 3.1.3.** Let X be a set and  $\mathcal{E} \subset \mathcal{P}(X)$ . Set  $\mathcal{S} = \{ \mathcal{T} \subset \mathcal{P}(X) : \mathcal{T} \text{ is a topology on } X \text{ and } \mathcal{E} \subset \mathcal{T} \}$  We define the **topology generated by**  $\mathcal{E}$  on X, denoted  $\tau(\mathcal{E})$ , by

$$\tau(\mathcal{E}) = \bigcap_{\mathcal{T} \in \mathcal{S}} \mathcal{T}$$

**Definition 3.1.4.** Let  $(X, \mathcal{T})$  be a topological space and  $U \subset X$ . Then U is said to be **open** if  $U \in \mathcal{T}$  and U is said to be **closed** if  $U^c$  is open.

**Definition 3.1.5.** Let X be a topological space and  $S, N \subset X$ . Then N is said to be a **neighborhood** of S if there exists  $U \subset X$  such that U is open and  $S \subset U \subset N$ . For  $S \in X$ , we denote the set of neighborhoods of S by  $\mathcal{N}_S$ .

**Definition 3.1.6.** Let X be a topological space and  $A \subset X$ . Set  $\mathcal{U}_A = \{U \subset X : U \subset A \text{ and } U \text{ is open}\}$  and  $\mathcal{C}_A = \{U \subset X : A \subset U \text{ and } U \text{ is closed}\}$ . We define the **interior of A**, denoted  $A^{\circ}$ , by

$$A^{\circ} = \bigcup_{U \in \mathcal{U}_A} U$$

We define the **closure of A**, denoted  $\overline{A}$ , by

$$\overline{A} = \bigcap_{U \in \mathcal{C}_A} U$$

**Definition 3.1.7.** Let X be a topological space and  $A \subset X$ . Then

- (1) A is open iff  $A = A^{\circ}$
- (2) A is closed iff  $A = \overline{A}$

Proof. Clear.  $\Box$ 

**Exercise 3.1.8.** Let X be a topological space and  $A \subset X$ . Then  $(A^{\circ})^{c} = \overline{A^{c}}$ .

Proof.  $\Box$ 

**Definition 3.1.9.** Let X be a topological space,  $A \subset X$  and  $x \in X$ . Then x is said to be a **limit point of** A if for each  $U \in \mathcal{N}_x$ ,

$$A \cap (U \setminus \{x\}) \neq \emptyset$$

We define  $A' = \{x \in A : x \text{ is a limit point of } A\}.$ 

**Exercise 3.1.10.** Let X be a topological space and  $A \subset X$ . Then  $\overline{A} = A \cup A'$ .

Proof.

**Exercise 3.1.11.** Let X be a topological space,  $A \subset X$  and  $x \in X$ . Then  $A \in \mathcal{N}_x$  iff  $x \in A^{\circ}$ .

*Proof.* Suppose that  $A \in \mathcal{N}_x$ . Then there exists  $U \subset X$  such that U is open and  $x \in U \subset A$ . By definition,  $U \subset A^{\circ}$ . Conversely, suppose that  $x \in A^{\circ}$ . Then by definition,  $A^{\circ} \in \mathbb{N}_x$ .

**Exercise 3.1.12.** Let X be a topological space and  $A \subset X$ . Then A is open iff for each  $x \in A$ , there exists  $U \in \mathcal{N}_x$  such that U is open and  $U \subset A$ .

*Proof.* Suppose that A is open. Let  $x \in A$ . Then  $A \in \mathcal{N}_x$ , A is open and  $A \subset A$ . Conversely, suppose that or each  $x \in A$ , there exists  $U_x \in \mathcal{N}_x$  such that U is open and  $U_x \subset A$ . Then

$$A = \bigcup_{x \in A} U_x$$

is open.  $\Box$ 

#### 3.2. Continuous Functions.

**Definition 3.2.1.** Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be topological spaces and  $f: X \to Y$ . Then f is said to be **continuous** if for each  $B \in \mathcal{B}$ ,  $f^{-1}(B) \in \mathcal{A}$ .

**Definition 3.2.2.** Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be topological spaces,  $f : X \to Y$  and  $x \in X$ . Then f is said to be **continuous at** x if for each  $V \in \mathcal{N}_{f(x)}$ , there exists  $U \in \mathcal{N}_x$  such that  $f(U) \subset V$ .

**Exercise 3.2.3.** Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be topological spaces,  $f: X \to Y$  and  $x \in X$ . Then f is continuous at x iff for each  $V \in \mathcal{N}_{f(x)}$ ,  $f^{-1}(V) \in \mathcal{N}_x$ .

*Proof.* Suppose that f is continuous at x. Let  $V \in \mathcal{N}_{f(x)}$ . Then there exists  $U \in \mathcal{N}_x$  such that  $f(U) \subset V$ . Thus

$$x \in U^{\circ}$$

$$\subset U$$

$$\subset f^{-1}(f(U))$$

$$\subset f^{-1}(V)$$

So  $f^{-1}(V) \in \mathcal{N}_x$ .

Conversely, suppose that for each  $V \in \mathcal{N}_{f(x)}$ ,  $f^{-1}(V) \in \mathcal{N}_x$ . Let  $V \in \mathcal{N}_{f(x)}$ . Hence  $f^{-1}(V) \in \mathcal{N}_x$ . Set  $U = f^{-1}(V)$ . Then

$$f(U) = f(f^{-1}(V))$$

$$\subset V$$

Thus f is continuous at x.

**Exercise 3.2.4.** Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be topological spaces and  $f: X \to Y$ . Then f is continuous iff for each  $x \in X$ , f is continuous at x.

Proof. Suppose that f is continuous. Let  $x \in X$ . Let  $V \in \mathcal{N}_{f(x)}$ . Then  $V^{\circ} \in \mathcal{B}$  and  $f(x) \in V^{\circ}$ . Set  $U = f^{-1}(V^{\circ})$ . By continuity,  $U \in \mathcal{A}$  and by construction,  $x \in U$ . Hence  $U \in \mathcal{N}_x$ . Then

$$f(U) = f(f^{-1}(V^{\circ}))$$

$$\subset V^{\circ}$$

$$\subset V$$

So f is continuous at x.

Conversely, suppose that for each  $x \in X$ , f is continuous at x. Let  $B \in \mathcal{B}$ . Let  $x \in f^{-1}(B)$ . Then  $B \in \mathcal{N}_{f(x)}$ . Continuity at x implies that  $f^{-1}(B) \in \mathcal{N}_x$ . Then  $x \in (f^{-1}(B))^{\circ}$ . Since  $x \in f^{-1}(B)$  is arbitrary,  $f^{-1}(B) \subset (f^{-1}(B))^{\circ}$ . Hence  $f^{-1}(B) = (f^{-1}(B))^{\circ}$  which implies that  $f^{-1}(B) \in \mathcal{A}$ . So f is continuous.  $\square$ 

**Definition 3.2.5.** Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be topological spaces and  $f: X \to Y$ . We define the

(1) push-forward of  $\mathcal{A}$ , denoted  $f_*\mathcal{A}$ , by

$$f_*\mathcal{A} = \{B \subset Y : f^{-1}(B) \in \mathcal{A}\}\$$

(2) pull-back of  $\mathcal{B}$ , denoted  $f^*\mathcal{B}$ , by

$$f^*\mathcal{B} = \{f^{-1}(B) : B \in \mathcal{B}\}$$

**Exercise 3.2.6.** Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be topological spaces and  $f: X \to Y$ . Then

- (1)  $f_*\mathcal{A}$  is a topology on Y
- (2)  $f^*\mathcal{B}$  is a topology on X

Proof.

(1) • Since  $f^{-1}(Y) = X \in \mathcal{A}$  and  $f^{-1}(\emptyset) = \emptyset \in \mathcal{A}, Y, \emptyset \in f_*\mathcal{A}$ .

• Let  $(U_{\alpha})_{\alpha \in A} \subset f_* \mathcal{A}$ . Then for each  $\alpha \in A$ ,  $f^{-1}(U_{\alpha}) \in \mathcal{A}$ . This implies that

$$f^{-1}\left(\bigcup_{\alpha\in A}U_{\alpha}\right) = \bigcup_{\alpha\in A}f^{-1}(U_{\alpha})$$
$$\in \mathcal{A}$$

Hence  $\bigcup_{\alpha \in A} U_{\alpha} \in f_* \mathcal{A}$ .

• Let  $(U_j)_{j=1}^n \subset f_* \mathcal{A}$ . Then for each  $j \in 1, \ldots, n, f^{-1}(U_j) \in \mathcal{A}$ . This implies that

$$f^{-1}\left(\bigcap_{j=1}^{n} U_{j}\right) = \bigcap_{j=1}^{n} f^{-1}(U_{j})$$

$$\in \mathcal{A}$$

Hence 
$$\bigcap_{j=1}^{n} U_j \in f_* \mathcal{A}$$
.

So  $f_*\mathcal{A}$  is a topology on Y.

(2) Similar to (1).

**Exercise 3.2.7.** Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be topological spaces,  $f: X \to Y$  and  $\mathcal{E} \subset \mathcal{P}(Y)$ . Suppose that  $\mathcal{B} = \tau(\mathcal{E})$ . Then f is continuous iff for each  $B \in \mathcal{E}$ ,  $f^{-1}(B) \in \mathcal{A}$ .

*Proof.* Suppose that f is continuous. Since  $\mathcal{E} \subset \mathcal{B}$ , clearly for each  $B \in \mathcal{E}$ ,  $f^{-1}(B) \in \mathcal{A}$ . Conversely, suppose that for each  $B \in \mathcal{E}$ ,  $f^{-1}(B) \in \mathcal{A}$ . Then  $\mathcal{E} \subset f_*\mathcal{A}$ . Since  $f_*\mathcal{A}$  is a topology on Y, we have that  $\mathcal{B} = \tau(\mathcal{E}) \subset f_*\mathcal{A}$ . So f is continuous.

**Definition 3.2.8.** Let X be a set,  $(Y_{\alpha}, \mathcal{B}_{\alpha})_{\alpha \in A}$  a collection of topological spaces and  $\mathcal{F} \in \prod_{\alpha \in A} Y_{\alpha}^{X}$  (i.e.  $\mathcal{F} = (f_{\alpha})_{\alpha \in A}$  where for each  $\alpha \in A$ ,  $f_{\alpha} : X \to Y_{\alpha}$ ). We define the **weak** 

topology generated by  $\mathcal{F}$  on X, denoted  $\tau(\mathcal{F})$ , by

$$\tau(\mathcal{F}) = \tau \left( \bigcup_{\alpha \in A} f_{\alpha}^* \mathcal{B}_{\alpha} \right)$$
$$= \tau(\{f_{\alpha}^{-1}(B) : B \in \mathcal{B}_{\alpha} \text{ and } \alpha \in A\})$$

**Definition 3.2.9.** Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be topological spaces and  $f: X \to Y$ . Then

- (1) f is said to be open if for each  $A \in \mathcal{A}$ ,  $f(A) \in \mathcal{B}$ .
- (2) f is said to be **closed** if for each  $A \subset X$ , if  $A^c \in \mathcal{A}$ , then  $f(A)^c \in \mathcal{B}$ .

3.3. Nets.

**Definition 3.3.1.** Let A be a set and  $\leq$  a relation on A. Then  $(A, \leq)$  is said to be a **directed set** if,

- (1) for each  $\alpha \in A$ ,  $\alpha \leq \alpha$
- (2) for each  $\alpha, \beta, \gamma \in A$ ,  $\alpha \leq \beta$  and  $\beta \leq \gamma$  implies that  $\alpha \leq \gamma$
- (3) for each  $\alpha, \beta \in A$ , there exists  $\gamma \in A$  such that  $\alpha, \beta \leq \gamma$

**Definition 3.3.2.** Let X be a topological space, A a directed set and  $x : A \to Y$ . Then x is said to be a **net** in X. We typically write  $(x_{\alpha})_{{\alpha} \in A}$ .

**Definition 3.3.3.** Let X be a topological space,  $(x_{\alpha})_{\alpha \in A} \subset X$  a net and  $U \subset X$ . Then  $(x_{\alpha})_{\alpha \in A}$  is said to be

- in U eventually if there exists  $\beta \in A$  such that for each  $\alpha \geq \beta$ ,  $x_{\alpha} \in U$
- in U infinitely often if for each  $\alpha \in A$ , there exists  $\beta \in A$  such that  $\beta \geq \alpha$  and  $x_{\beta} \in U$

**Definition 3.3.4.** Let X be a topological space,  $(x_{\alpha})_{\alpha \in A} \subset X$  a net and  $x \in X$ . Then  $(x_{\alpha})_{\alpha \in A}$  is said to **converge to** x, denoted  $x_{\alpha} \to x$ , if for each  $U \in \mathcal{N}_x$ ,  $(x_{\alpha})_{\alpha \in A}$  is eventually in U.

**Definition 3.3.5.** Let X be a topological space and  $(x_{\alpha})_{\alpha \in A} \subset X$  a net. Then  $(x_{\alpha})_{\alpha \in A}$  is said to **converge** if there exists  $x \in X$  such that  $x_{\alpha} \to x$ .

**Exercise 3.3.6.** Let X be a topological space,  $A \subset X$  and  $x \in X$ . Then  $x \in \overline{A}$  iff there exists a net  $(x_{\alpha})_{\alpha \in A} \subset A$  such that  $x_{\alpha} \to x$ .

*Proof.* Suppose that  $x \in \overline{A}$ . If  $x \in A$ ,

**Exercise 3.3.7.** Let  $(X, \mathcal{A})$  and  $(Y, \mathcal{B})$  be topological spaces,  $f: X \to Y$  and  $x \in X$ . Then f is continuous at x iff for each net  $(x_{\alpha})_{\alpha \in A} \subset X$ ,  $x_{\alpha} \to x$  implies that  $f(x_{\alpha}) \to f(x)$ .

Proof. Suppose that f is continuous at x. Let  $(x_{\alpha})_{\alpha \in A} \subset X$  be a net. Suppose that  $x_{\alpha} \to x$ . Let  $V \in \mathcal{N}_{f(x)}$ . Continuity implies that  $f^{-1}(V) \in \mathcal{N}_x$ . Since  $x_{\alpha \to x}$ ,  $(x_{\alpha})_{\alpha \in A}$  is eventually in  $f^{-1}(V)$ . So there exists  $\beta \in A$  such that for each  $\alpha \in A$ ,  $\alpha \geq \beta$  implies that  $x_{\alpha} \in f^{-1}(V)$ . Let  $\alpha \in A$ . Suppose that  $\alpha \geq \beta$ . Then  $f(x_{\alpha}) \in V$ . So  $(f(x_{\alpha}))_{\alpha \in A}$  is eventually in V. Since  $V \in \mathcal{N}_{f(x)}$  is arbitrary,  $f(x_{\alpha}) \to f(x)$ .

Conversely, suppose that f is not continuous at x. Then there exists  $V \in \mathcal{N}_{f(x)}$  such that  $f^{-1}(V) \notin \mathcal{N}_x$ . Then  $x \notin (f^{-1}(V))^{\circ}$ . So  $x \in ((f^{-1}(V))^{\circ})^c = \overline{f^{-1}(V^c)}$ . This implies that there exists a net  $(x_{\alpha})_{\alpha \in A} \subset f^{-1}(V^c)$  such that  $x_{\alpha} \to x$ .

for each net 
$$(x_{\alpha})_{\alpha \in A} \subset X$$
,  $x_{\alpha} \to x$  implies that  $f(x_{\alpha}) \to f(x)$ .

Exercise 3.3.8.

Exercise 3.3.9.

**Exercise 3.3.10.** Let X, Y be topological spaces and  $\phi: X \to Y$  a homeomorphism. Then for each  $E \subset X$ ,

- $(1) \ \overline{\phi(E)} = \phi(\overline{E})$
- (2)  $\phi(E)^{\circ} = \phi(E^{\circ})$

Proof.

(1) Let  $E \subset X$ . Since  $\overline{E} \subset \overline{E}$ , we have that  $\phi(E) \subset \phi(\overline{E})$ . Since  $\overline{E}$  is closed,  $\phi(\overline{E})$  is closed and thus  $\overline{\phi(E)} \subset \phi(\overline{E})$ . Conversely, let  $x \in \phi(\overline{E})$ . Then  $\phi^{-1}(x) \in \overline{E}$ . Then there exists a net  $(y_{\alpha})_{\alpha \in A} \subset E$  such that  $y_{\alpha} \to \phi^{-1}(x)$ . Then  $(\phi(y_{\alpha}))_{\alpha \in A} \subset \phi(E)$  and  $\phi(y_{\alpha}) \to x$ . Thus  $x \in \overline{\phi(E)}$  and  $\phi(\overline{E}) \subset \overline{\phi(E)}$ .

(2) Similar

Definition 3.3.11.

Exercise 3.3.12.

#### 3.4. Semi-continuity.

**Definition 3.4.1.** Let X be a topological space,  $f: X \to (\infty, \infty]$  and  $x_0 \in X$ . Then f is said to be **lower semicontinuous (l.s.c.) at**  $x_0$  if

$$\liminf_{x \to x_0} f(x) \ge f(x_0)$$

and f is said to be **lower semicontinuous** (l.s.c.) if for each  $x_0 \in X$ , f is lower semicontinuous at  $x_0$ .

**Exercise 3.4.2.** Let X be a topological space and  $f: X \to (\infty, \infty]$ . Then f is l.s.c. iff for each  $\alpha \in \mathbb{R}$ ,  $f^{-1}((\alpha, \infty])$  is open.

*Proof.* Suppose that f is l.s.c. Let  $\alpha \in \mathbb{R}$  and  $x_0 \in f^{-1}(\alpha, \infty]$ . Put  $\epsilon = f(x_0) - \alpha$ . By definition,

$$\sup_{V \in N_{x_0}} \inf_{x \in V \setminus \{x_0\}} f(x) \ge f(x_0)$$

Choose  $V_{\epsilon} \in N_{x_0}$  such that

$$\inf_{x \in V_{\epsilon}} f(x) > f(x_0) - \epsilon$$
$$= \alpha$$

Then  $V_{\epsilon}^{o} \in \mathcal{N}_{x_0}$  is open and

$$V_{\epsilon}^{o} \subset V_{\epsilon}$$
  
 $\subset f^{-1}((\alpha, \infty])$ 

So  $f^{-1}((\alpha,\infty])$  is open.

Conversely, suppose that for each  $\alpha \in \mathbb{R}$ ,  $f^{-1}((\alpha, \infty])$  is open. Let  $x_0 \in X$ . Put  $\alpha = f(x_0)$ . For  $n \in \mathbb{N}$ , define  $V_n = f^{-1}((f(x_0) - 1/n, \infty])$ . Then for each  $n \in \mathbb{N}$ ,  $V_n \in \mathcal{N}_{x_0}$  and

$$\lim_{x \to x_0} \inf f(x) = \sup_{V \in \mathcal{N}_{x_0}} \inf_{x \in V \setminus \{x_0\}} f(x)$$

$$\geq \sup_{n \in \mathbb{N}} \inf_{x \in V_n \setminus \{x_0\}} f(x)$$

$$\geq \sup_{n \in \mathbb{N}} f(x_0) - 1/n$$

$$= f(x_0)$$

So f is l.s.c.

#### 4. Banach Spaces

#### 4.1. Introduction.

**Note 4.1.1.** In the following, we will consider vector spaces over  $\mathbb{C}$ . There are analogous results for real vector spaces as well, just replace every  $\mathbb{C}$  with  $\mathbb{R}$ .

**Definition 4.1.2.** Let X be a normed vector space. Then X is said to be a **Banach space** if X is complete.

**Definition 4.1.3.** Let X be a normed vector space and  $(x_i)_{i=1}^n \subset X$ . The series  $\sum_{i=1}^\infty x_i$  is said to **converge** if the sequence  $s_n := \sum_{i=1}^n x_i$  converges. The series  $\sum_{i=1}^\infty x_i$  is said to **converge absolutely** if  $\sum_{i\in\mathbb{N}} ||x_i|| < \infty$ .

**Exercise 4.1.4.** Let X be a normed vector space. Then X is complete iff for each  $(x_i)_{i\in\mathbb{N}}\subset X$ ,  $\sum_{i=1}^{\infty}x_i$  converges absolutely implies that  $\sum_{i=1}^{\infty}x_i$  converges.

**Hint:** Given a Cauchy sequence  $(x_n)_{n\in\mathbb{N}}$ , obtain a subsequence  $(x_{n_j})_{j\in\mathbb{N}}\subset (x_n)_{n\in\mathbb{N}}$  such that for each  $j\in\mathbb{N}$ ,  $||x_{n_{j+1}}-x_{n_j}||<2^{-j}$ . Define a new sequence  $(y_j)_{j\in\mathbb{N}}\subset X$  by

$$y_j = \begin{cases} x_{n_1} & j = 1\\ x_{n_j} - x_{n_{j-1}} & j \ge 2 \end{cases}$$

*Proof.* Suppose that X is complete. Let  $(x_i)_{i \in \mathbb{N}} \subset X$ . Suppose that  $\sum_{i=1}^{\infty} x_i$  converges absolutely. Let  $\epsilon > 0$ . Choose  $N \in \mathbb{N}$  such that for each  $m, n \in \mathbb{N}$ , if  $m, n \geq N$  and m < n, then  $\sum_{m+1}^{n} \|x_i\| < \epsilon$ . Let  $m, n \in \mathbb{N}$ . Suppose that m < n. Then

$$||s_n - s_m|| = \left\| \sum_{i=1}^n x_i - \sum_{i=1}^m x_i \right\|$$

$$= \left\| \sum_{i=m+1}^n x_i \right\|$$

$$\leq \sum_{i=m+1}^n ||x_i||$$

$$\leq \epsilon$$

Thus  $(s_n)_{n\in\mathbb{N}}$  is Cauchy. Since X is complete,  $\sum_{i=1}^{\infty}x_i$  converges. Conversely, Suppose that for each  $(x_i)_{i\in\mathbb{N}}\subset X$ ,  $\sum_{i=1}^{\infty}x_i$  converges absolutely implies that  $\sum_{i=1}^{\infty}x_i$  converges. Let  $(x_i)_{i\in\mathbb{N}}\subset X$  be Cauchy. Proceed inductively to create a strictly increasing sequence  $(n_i)_{i\in\mathbb{N}}\subset\mathbb{N}$  such that for each  $m,n\in\mathbb{N}$ , if  $m,n\geq n_i$ , then  $||x_m-x_n||<2^{-i}$ . Define  $(y_i)_{i\in\mathbb{N}}\subset X$  by

$$y_i = \begin{cases} x_{n_1} & i = 1\\ x_{n_i} - x_{n_{i-1}} & i \ge 2 \end{cases}$$

Then  $\sum_{i=1}^{k} y_i = x_{n_k}$  and

$$\sum_{i \in \mathbb{N}} ||y_i|| = ||x_{n_1}|| + \sum_{i \in \mathbb{N}} ||x_{n_i} - x_{n_{i-1}}||$$

$$\leq ||x_{n_1}|| + 2 \sum_{i \in \mathbb{N}} 2^{-i}$$

$$= ||x_{n_i}|| + 2$$

Hence  $(x_{n_k})_{k\in\mathbb{N}} = (\sum_{i=1}^k y_i)_{i\in\mathbb{N}}$  converges. Since  $(x_i)_{i\in\mathbb{N}}$  is cauchy and has a convergent subsequence, it converges. So X is complete.

**Exercise 4.1.5.** Let X be a normed vector space. Then addition and scalar multiplication are continuous on  $X \times X$  and  $\|\cdot\|: X \to [0, \infty)$  is continuous.

*Proof.* Let 
$$\epsilon > 0$$
. Choose  $\delta = \frac{\epsilon}{2}$ . Let  $(x_1, y_1), (x_2, y_2) \in X \times X$ . Suppose that  $\max\{\|x_1 - x_2\|, \|y_1 - y_2\|\} < \delta$ 

Then

$$||(x_1 + y_1) - (x_2 + y_2)|| = ||(x_1 - x_2) + (y_1 - y_2)||$$

$$\leq ||x_1 - x_2|| + ||y_1 - y_2||$$

$$< 2\delta$$

$$= \epsilon$$

Hence addition is uniformly continuous.

Let  $(\lambda_1, x_1) \in \mathbb{C} \times X$  and  $\epsilon > 0$ . Choose  $\delta = \min\{\frac{\epsilon}{2(|\lambda_1| + ||x_1|| + 1)}, \frac{\sqrt{\epsilon}}{\sqrt{2}}\}$ . Let  $(\lambda_2, x_2) \in \mathbb{C} \times X$ . Suppose that

$$\max\{|\lambda_1 - \lambda_2|, ||x_1 - x_2||\} < \delta$$

Then

$$\|\lambda_{1}x_{1} - \lambda_{2}x_{2}\| = \|\lambda_{1}x_{1} - \lambda_{1}x_{2} + \lambda_{1}x_{2} - \lambda_{2}x_{2}\|$$

$$= \|\lambda_{1}(x_{1} - x_{2}) + (\lambda_{1} - \lambda_{2})x_{2}\|$$

$$\leq |\lambda_{1}| \|x_{1} - x_{2}\| + |\lambda_{1} - \lambda_{2}| \|x_{2}\|$$

$$\leq |\lambda_{1}| \|x_{1} - x_{2}\| + |\lambda_{1} - \lambda_{2}| (\|x_{1} - x_{2}\| + \|x_{1}\|)$$

$$< |\lambda_{1}|\delta + \delta(\delta + \|x_{1}\|)$$

$$= (|\lambda_{1}| + \|x_{1}\|)\delta + \delta^{2}$$

$$< \frac{\epsilon}{2} + \frac{\epsilon}{2}$$

$$= \epsilon$$

Since  $(\lambda_1, x_1) \in \mathbb{C} \times X$  is arbitrary, scalar multiplication is continuous.

Let  $\epsilon > 0$ . Choose  $\delta = \epsilon$ . Let  $x, y \in X$ . Suppose that  $||x - y|| < \delta$ . Then

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

$$< \delta$$

$$= \epsilon$$

So  $\|\cdot\|:X\to [0,\infty)$  is uniformly continuous.

#### 4.2. Linear Maps.

**Definition 4.2.1.** Let X, Y be a normed vector spaces. A linear map  $T: X \to Y$  is said to be **bounded** if there exists  $C \ge 0$  such that for each  $x \in X$ ,

$$||Tx|| \le C||x||$$

We define

$$L(X,Y) = \{T : X \to Y : T \text{ is linear and bounded}\}$$

When X = Y, we write L(X).

**Exercise 4.2.2.** Set  $X = C^1([0,1])$  and Y = C([0,1]). Equip both X and Y with the sup norm. Define  $T: X \to Y$  by Tf = f'. Then T is not bounded.

*Proof.* For the sake of contradiction, suppose that T is bounded. Then there exists  $C \geq 0$  such that for each  $f \in X$ ,  $||Tf|| \leq C||f||$ . Choose  $n \in \mathbb{N}$  such that n > C. Define  $f \in X$  by  $f(x) = x^n$ . Then

$$n = ||Tf||$$

$$\leq C||f||$$

$$= C$$

which is a contradiction. Hence T is not bounded.

**Exercise 4.2.3.** Let X, Y be a normed vector spaces and  $T: X \to Y$  a linear map. Then T is bounded iff there exists r, s > 0 such that  $T(B(0, r)) \subset B(0, s)$ 

*Proof.* Suppose that T is bounded. Then there exists  $C \geq 0$  such that for each  $x \in X$ ,  $||Tx|| \leq C||x||$ . Thus  $T(B(0,1)) \subset B(0,C+1)$ . Conversely. Suppose that there exists r,s>0 such that  $T(B(0,r)) \subset B(0,s)$ . Define  $C=\frac{2s}{r}$ . Let  $x \in X$ . Put  $\alpha=\frac{r}{2||x||}$  Then  $\alpha x \in B(0,r)$ . So  $T(\alpha x) = \alpha T(x) \in B(0,s)$ . Hence

$$||T(\alpha x)|| = ||\alpha T(x)||$$

$$= |\alpha||T(x)||$$

$$= \frac{r}{2||x||}||T(x)||$$

$$< s.$$

Thus

$$||Tx|| < \frac{2s}{r}||x|| = C||x||$$

So T is bounded.

**Exercise 4.2.4.** Let X, Y be normed vector spaces and  $T: X \to Y$ . Suppose that T is linear. Then there exists  $x_0 \in X$  such that T is continuous at  $x_0$  iff T is continuous at 0.

*Proof.* Suppose that there exists  $x_0 \in X$  such that T is continuous at  $x_0$ . Since T is linear, T(0) = 0. Let  $(x_n)_{n \in \mathbb{N}} \subset X$ . Suppose that  $x_n \to 0$ . Then  $x_n + x_0 \to x_0$ . Hence

$$T(x_n) + T(x_0) = T(x_n + x_0)$$
$$\to T(x_0)$$

This implies that

$$T(x_n) \to 0$$
$$= T(0)$$

Therefore T is continuous at 0.

Conversely, if T is continuous at 0, then trivially, there exists  $x_0 \in X$  such that T is continuous at  $x_0$ .

**Exercise 4.2.5.** Let X, Y be normed vector spaces and  $T: X \to Y$  a linear map. Then the following are equivalent:

- (1) T is continuous
- (2) T is continuous at x = 0
- (3) T is bounded

Proof.

- $(1) \Longrightarrow (2)$ : Trivial
- $\bullet$  (2)  $\Longrightarrow$  (3):

Suppose that T is continuous at x=0. Then there exists  $\delta>0$  such that for each  $x\in X$ , if  $\|x\|<\delta$ , then  $\|Tx\|<1$ . Choose  $C=\frac{2}{\delta}$ . If x=0, then  $\|Tx\|\leq C\|x\|$ . Suppose that  $\|x\|\neq 0$ . Define  $y=\frac{\delta}{2\|x\|}x$ . Then  $\|y\|<\delta$ . So

$$1 > ||Ty||$$

$$= \frac{\delta}{2||x||}||Tx||$$

Thus

$$||Tx|| < \frac{2}{\delta}||x||$$
$$= C||x||$$

Hence T is bounded.

• (3)  $\Longrightarrow$  (1) Suppose that T is bounded. Then there exists  $C \ge 0$  such that for each  $x \in X$ ,  $||Tx|| \le C||x||$ . Let  $\epsilon > 0$ . Choose  $\delta = \frac{\epsilon}{C+1}$ . Let  $x, y \in X$  Suppose that  $||x - y|| < \delta$ . Then

$$||Tx - Ty|| = ||T(x - y)||$$

$$\leq C||x - y||$$

$$< (C + 1)\delta$$

$$= \epsilon$$

So T is continuous.

**Definition 4.2.6.** Let X, Y be normed vector spaces. Define  $\|\cdot\| : L(X, Y) \to [0, \infty)$  by  $\|T\| = \inf\{C \ge 0 : \text{for each } x \in X, \|Tx\| \le C\|x\|\}$ 

We call  $\|\cdot\|$  the **operator norm on** L(X,Y)

**Exercise 4.2.7.** Let X, Y be normed vector spaces. If  $X \neq \{0\}$ , then the operator norm on L(X,Y) is given by:

- (1)  $||T|| = \sup ||Tx||$
- (2)  $||T|| = \sup_{x \neq 0} ||x||^{-1} ||Tx||$
- (3)  $||T|| = \inf\{C \ge 0 : \text{for each } x \in X, \, ||Tx|| \le C||x||\}$

*Proof.* Since  $X \neq \{0\}$ , the supremums in (1) and (2) are well defined. Let  $T \in L(X,Y)$ . By linearity of T, the sets over which the supremums are taken in (1) and (2) are the same. So (1) and (2) are equal.

Now, set  $M = \sup \|Tx\|$  and  $m = \inf\{C \ge 0 : \text{ for each } x \in X, \|Tx\| \le C\|x\|\}$ . Let  $x \in X$ .

If ||x|| = 0, then  $||Tx|| \le M||x||$ . Suppose that  $||x|| \ne 0$ . Then

$$||Tx|| = \left( ||T(x/||x||)|| \right) ||x||$$

$$\leq M||x||$$

Hence  $M \in \{C \geq 0 : \text{ for each } x \in X, \|Tx\| \leq C\|x\|\}$  and  $m \leq M$ . Let  $C \in \{C \geq 0 : A\}$ for each  $x \in X$ ,  $||Tx|| \le C||x||$ . Suppose that ||x|| = 1. Then  $||Tx|| \le C||x|| = C$ . So  $M \leq C$ . Therefore  $M \leq m$ . So M = m and the supremum in (1) is the same as the infimum in (3). 

Note 4.2.8. From here on, unless stated otherwise, we assume  $X \neq 0$ .

**Exercise 4.2.9.** Let X, Y be normed vector spaces and  $T \in L(X,Y)$ . Then for each  $x \in X$ ,  $||Tx|| \le ||T|| ||x||$ 

*Proof.* This is just part of the previous exercise. Let  $x \in X$ . If x = 0, then  $||Tx|| \le ||T|| ||x||$ . Suppose that  $x \neq 0$ . Then  $||Tx|| = T(x/||x||)||x|| \leq ||T||||x||$ 

**Exercise 4.2.10.** Let X, Y be normed vector spaces. Then the operator norm is a norm on L(X,Y).

*Proof.* Let  $S, T \in L(X, Y)$  and  $\alpha \in \mathbb{C}$ . For each  $x \in X$ , we have that

$$||(S+T)x|| = ||Sx + Tx||$$

$$\leq ||Sx|| + ||Tx||$$

$$\leq ||S|| ||x|| + ||T|| ||x||$$

$$= (||S|| + ||T||) ||x||$$

So ||S + T|| < ||S|| + ||T||.

Using the definition of ||T||, we see that

$$\|\alpha T\| = \sup_{\|x\|=1} \|(\alpha T)x\|$$

$$= \sup_{\|x\|=1} |\alpha| \|Tx\|$$

$$= |\alpha| \sup_{\|x\|=1} \|Tx\|$$

$$= |\alpha| \|T\|$$

So  $\|\alpha S\| = |\alpha| \|S\|$ .

Suppose that ||T|| = 0. Let  $x \in X$ . Then  $||Tx|| \le ||T|| ||x|| = 0$ . So Tx = 0. Since  $x \in X$  is arbitrary, we have that T = 0.

**Exercise 4.2.11.** Let X, Y, Z be normed vector spaces,  $T \in L(X, Y)$  and  $S \in L(Y, Z)$ . Define  $ST : X \to Z$  by STx = S(Tx). Then  $ST \in L(X, Z)$  and  $||ST|| \le ||S|| ||T||$ .

*Proof.* Clearly ST is linear. Let  $x \in X$ . Then

$$||STx|| = ||S(Tx)||$$
  
 $\leq ||S|| ||Tx||$   
 $\leq ||S|| ||T|| ||x||$ 

So  $||ST|| \le ||S|| ||T||$ .

**Definition 4.2.12.** Let X, Y be a normed vector spaces and  $T \in L(X, Y)$ . Then T is said to be **invertible** or an **isomorphism** if T is a bijection and  $T^{-1} \in L(Y, X)$ .

**Definition 4.2.13.** Let X be a normed vector space. Define  $GL(X) := \{T \in L(X, X) : T \text{ is invertible}\}.$ 

**Exercise 4.2.14.** Let X, Y be normed vector spaces. If Y is complete, then so is L(X, Y).

Proof. Suppose that Y is complete. Let  $(T_n)_{n\in\mathbb{N}}\subset L(X,Y)$ . Suppose that  $(T_n)_{n\in\mathbb{N}}$  is Cauchy. Since for each  $m,n\in\mathbb{N}$ ,  $|\|T_m\|-\|T_n\||\leq \|T_m-T_n\|$ , we have that  $(\|T_n\|)_{n\in\mathbb{N}}\subset[0,\infty)$  is Cauchy. Hence  $\lim_{n\to\infty}\|T_n\|$  exists.

Let  $x \in X$  and  $m, n \in \mathbb{N}$ . Then

$$||T_m x - T_n x|| = ||(T_m - T_n)x||$$
  
 $\leq ||T_m - T_n|| ||x||$ 

So  $(T_n x)_{n \in \mathbb{N}} \subset Y$  is Cauchy and hence converges. Define  $T: X \to Y$  by  $Tx = \lim_{n \to \infty} T_n x$ .

Since addition and scalar multiplication are continuous, T is linear. Let  $x \in X$  and  $\epsilon > 0$ . Choose  $N \in \mathbb{N}$  such that for each  $n \in N$ , if  $n \geq N$ , then  $||Tx - T_nx|| < \epsilon$ . Then for each

 $n \in \mathbb{N}$ , if  $n \geq N$  we have that

$$||Tx|| \le ||Tx - T_n x|| + ||T_n x||$$

$$< \epsilon + ||T_n x||$$

$$\le \epsilon + ||T_n|| ||x||$$

Thus  $||Tx|| \le \epsilon + (\lim_{n \to \infty} ||T_n||)||x||$ . Since  $\epsilon > 0$  is arbitrary,  $||Tx|| \le (\lim_{n \to \infty} ||T_n||)||x||$ . Thus  $T \in L(X, Y)$  and  $||T|| \le \lim_{n \to \infty} ||T_n||$ .

Note that since addition, scalar multiplication and  $\|\cdot\|$  are continuous, we have that for each  $n \in \mathbb{N}$  and  $x \in X$ ,  $\|(T_n - T_m)x\|$  converges to  $\|(T_n - T)x\|$  because

$$\lim_{m \to \infty} \|(T_n - T_m)x\| = \lim_{m \to \infty} \|T_n x - T_m x\|$$

$$= \|T_n x - \lim_{m \to \infty} T_m x\|$$

$$= \|T_n x - Tx\|$$

$$= \|(T_n - T)x\|$$

Let  $\epsilon > 0$ . Choose  $N \in \mathbb{N}$  such that for each  $m, n \in \mathbb{N}$  if  $n, m \geq N$ , then  $||T_n - T_m|| < \epsilon$ . Then for each  $n \in \mathbb{N}$  if  $n \geq N$ , then for each  $x \in X$ ,

$$||(T_n - T_m)x|| \le ||(T_n - T_m)||||x|| < \epsilon ||x||$$

Combining this with the previous fact, we see that for each  $n \in \mathbb{N}$ , if  $n \geq \mathbb{N}$ , then for each  $x \in X$ ,

$$||(T_n - T)x|| \le \epsilon ||x||$$

In particular, for each  $n \in \mathbb{N}$ , if  $n \geq N$ , then

$$||T_n - T|| = \sup_{\|x\|=1} ||(T_n - T)x|| \le \epsilon$$

This implies that  $T_n$  converges to T in L(X,Y). Since

$$||T_n|| - ||T||| \le ||T_n - T||$$

it is clear that  $\lim_{n\to\infty} ||T_n|| = ||T||$ 

#### 4.3. Direct Sums.

**Definition 4.3.1.** Let X, Y be normed vector spaces and  $p \in [1, \infty]$ . Let  $\|\cdot\|_p' : \mathbb{R}^2 \to [0, \infty)$  denote the usual  $l^p$  norm. We define  $\|\cdot\|_p : X \oplus Y \to [0, \infty)$  by

$$||(x,y)||_p = ||(||x||, ||y||)||'_p$$

**Exercise 4.3.2.** Let X, Y be normed vector spaces. Then

- (1) for each  $p \in [1, \infty]$ ,  $\|\cdot\|_p : X \oplus Y \to [0, \infty)$  is a norm on  $X \oplus Y$
- (2)  $\{\|\cdot\|_p : p \in [1,\infty]\}$  are equivalent.

Proof.

- (1) Let  $p \in [1, \infty]$ ,  $(x_1, y_1)$ ,  $(x_2, y_2) \in X \oplus Y$  and  $\lambda \in \mathbb{C}$ .
  - Clearly if  $(x_1, y_1) = (0, 0)$ , then  $||S||_p = 0$ . Conversely, suppose that  $||(x_1, y_1)||_p = 0$ . Then  $||x_1|| = 0$  and  $||y_1|| = 0$ . So  $x_1 = 0$  and  $y_1 = 0$ . Therefore S = 0.

$$\|\lambda(x_1, y_1)\|_p = \|(\|\lambda x_1\|, \|\lambda y_1\|)\|_p'$$

$$= \|(|\lambda| \|x_1\|, |\lambda| \|y_1\|)\|_p'$$

$$= \||\lambda| (\|x_1\|, \|y_1\|)\|_p'$$

$$= |\lambda| \|(\|x_1\|, \|y_1\|)\|_p'$$

$$= |\lambda| \|(x_1, y_1)\|_p$$

 $\begin{aligned} \|(x_1, y_1) + (x_2, y_2)\|_p &= \|(\|x_1 + x_2\|, \|y_1 + y_2\|)\|_p' \\ &\leq \|(\|x_1\| + \|x_2\|, \|y_1\| + \|y_2\|)\|_p' \\ &= \|(\|x_1\|, \|y_1\|) + (\|x_2\|, \|y_2\|)\|_p' \\ &\leq \|(\|x_1\|, \|y_1\|)\|_p' + \|(\|x_2\|, \|y_2\|)\|_p' \\ &= \|(x_1, y_1)\|_p + \|(x_2, y_2)\|_p \end{aligned}$ 

(2) All norms on  $\mathbb{R}^2$  are equivalent.

**Exercise 4.3.3.** Let X, Y be Banach spaces. Then  $X \oplus Y$  equipped with  $\|\cdot\|_p : X \oplus Y \to [0, \infty)$  is a Banach space.

Proof. 
$$\Box$$

**Exercise 4.3.4.** Let X, Y and Z be Banach spaces and  $p \in [0, \infty]$ . Equip  $Y \oplus Z$  with  $\| \cdot \|_p$ . Let  $T \in L(X, Y \oplus Z)$  with  $T = (T_Y, T_Z)$ . Then  $T_Y \in L(X, Y)$  and  $T_Z \in L(X, Z)$ .

*Proof.* Let 
$$x \in X$$
. Then  $||T_Y(x)||, ||T_Z(x)|| \le FINISH!!!$ 

**Definition 4.3.5.** Let X, Y and Z be Banach spaces and  $p \in [0, \infty]$ . Let  $\|\cdot\|_p' : \mathbb{R}^2 \to [0, \infty)$  denote the usual  $l^p$  norm. Equip  $Y \oplus Z$  with  $\|\cdot\|_p$ . Let  $T \in L(X, Y \oplus Z)$  with  $T = (T_Y, T_Z)$ . Define  $\|\cdot\|_p : L(X, Y \oplus Z) \to [0, \infty)$  by

$$||T||_p = ||(||T_Y||, ||T_Z||)||_p'$$

**Exercise 4.3.6.** Let X, Y and Z be Banach spaces and  $p \in [0, \infty]$ . Then  $\|\cdot\|_p : L(X, Y \oplus Z) \to [0, \infty)$  is a norm on  $L(X, Y \oplus Z)$ .

*Proof.* Let  $\lambda \in \mathbb{C}$  and  $S, T \in L(X, Y \oplus Z)$  with  $S = (S_Y, S_Z)$  and  $T = (T_Y, T_Z)$ .

• Clearly if S = 0, then  $||S||_p = 0$ . Conversely, suppose that  $||S||_p = 0$ . Then  $||S_Y|| = 0$  and  $||S_Z|| = 0$ . So  $S_Y = 0$  and  $S_Z = 0$ . Therefore S = 0.

$$\|\lambda S\|_{p} = \|(\|\lambda S_{Y}\|, \|\lambda S_{Z}\|)\|'_{p}$$

$$= \|(|\lambda|\|S_{Y}\|, |\lambda|\|S_{Z}\|)\|'_{p}$$

$$= \||\lambda|(\|S_{Y}\|, \|S_{Z}\|)\|'_{p}$$

$$= |\lambda|\|(\|S_{Y}\|, \|S_{Z}\|)\|'_{p}$$

$$= |\lambda|\|S\|_{p}$$

•

$$||S + T||_{p} = ||(||S_{Y} + T_{Y}||, ||S_{Z} + T_{Z}||)||'_{p}$$

$$\leq ||(||S_{Y}|| + ||T_{Y}||, ||S_{Z}|| + ||T_{Z}||)||'_{p}$$

$$= ||(||S_{Y}||, ||S_{Z}||) + (||T_{Y}||, ||T_{Z}||)||'_{p}$$

$$\leq ||(||S_{Y}||, ||S_{Z}||)||'_{p} + ||(||T_{Y}||, ||T_{Y}||)||'_{p}$$

$$= ||S||_{p} + ||T||_{p}$$

So  $\|\cdot\|_p: L(X,Y\oplus Z)\to [0,\infty)$  is a norm on  $L(X,Y\oplus Z)$ .

**Exercise 4.3.7.** Let X, Y and Z be Banach spaces and  $p \in [0, \infty]$ . Equip  $Y \oplus Z$  with  $\|\cdot\|_p$ . Let  $T \in L(X, Y \oplus Z)$  with  $T = (T_Y, T_Z)$ . Then  $\|T\| \le 2^{1/p} \|T\|_p$ .

*Proof.* Let  $x \in X$ . If  $p < \infty$ , then

$$||T(x)||_{p} = ||(T_{Y}(x), T_{Z}(x))||_{p}$$

$$||(||T_{Y}(x)||, ||T_{Z}(x)||)||'_{p}$$

$$= \left(||T_{Y}(x)||^{p} + ||T_{Z}(x)||^{p}\right)^{1/p}$$

$$\leq \left(||T_{Y}||^{p}||x||^{p} + ||T_{Z}||^{p}||x||^{p}\right)^{1/p}$$

$$\leq \left[(||T_{Y}||^{p} + ||T_{Z}||^{p})||x||^{p} + (||T_{Y}||^{p} + ||T_{Z}||^{p})||x||^{p}\right]^{1/p}$$

$$= \left[2(||T_{Y}||^{p} + ||T_{Z}||^{p})||x||^{p}\right]^{1/p}$$

$$= 2^{1/p}||T||_{p}||x||$$

Hence 
$$||T|| \le 2^{1/p} ||T||_p$$
 If  $p = \infty$ , then 
$$||T(x)||_\infty = \max(||T_Y(x)||, ||T_Z(x)||)$$
 
$$\le \max(||T_Y|| ||x||, ||T_Z|| ||x||)$$
 
$$\le \max\left[\max(||T_Y||, ||T_Z||) ||x||, \max(||T_Y||, ||T_Z||) ||x||\right]$$
 
$$= \max(||T_Y||, ||T_Z||) ||x||$$
 
$$= ||T||_\infty ||x||$$

Hence

$$||T|| \le ||T||_{\infty}$$
$$= 2^{1/\infty} ||T||_{\infty}$$

**Exercise 4.3.8.** Let X and  $X_1, \dots, X_n$  be Banach spaces and  $p \in [0, \infty]$ . Equip  $\bigoplus_{j=1}^n X_j$  with  $\|\cdot\|_p$ . Let  $T \in L(X, \bigoplus_{j=1}^n X_j)$ . Then  $\|T\| \le n^{1/p} \|T\|_p$ .

*Proof.* Similar to the previous exercise.

#### 4.4. Quotient Spaces.

**Definition 4.4.1.** Let X be a normed vector space and  $M \subset X$  a closed subspace. Define  $\|\cdot\|: X/M \to [0,\infty)$  by

$$||x + M|| := \inf_{y \in M} ||x + y||$$

We call  $\|\cdot\|$  the subspace norm on X/M

**Exercise 4.4.2.** Let X be a normed vector space and  $M \subsetneq X$  a proper, closed subspace of M. Then

- (1) The previously defined subspace norm on X/M is well defined and is a norm.
- (2) For each  $\epsilon > 0$ , there exists  $x \in X$  such that ||x|| = 1 and  $||x + M|| \ge 1 \epsilon$ .
- (3) The projection map  $\pi: X \to X/M$  defined by  $\pi(x) = x + M$  is continuous and  $\|\pi\| = 1$ .
- (4) If X is complete, then X/M is complete.

Proof.

(1) Let  $x, y \in X$  and  $\alpha \in \mathbb{C}$ . Suppose that x+M=y+M. Then there exists  $m \in M$  such that x=y+m. Since M is a subspace, the map  $T:M\to M$  given by Tx=x+m is a bijection. So

$$\inf_{z \in M} \|y + m + z\| = \inf_{z \in M} \|y + z\|$$

which implies that

$$\begin{split} \|x + M\| &= \inf_{z \in M} \|x + z\| \\ &= \inf_{z \in M} \|y + m + z\| \\ &= \inf_{z \in M} \|y + z\| \\ &= \|y + M\| \end{split}$$

So  $\|\cdot\|: X/M \to [0,\infty)$  is well defined.

We observe that for each  $z, w \in M$ ,

$$||x + y + z|| \le ||x + w|| + ||y + w + z||$$

Taking infimums over M with respect to z in this inequality implies that for each  $w \in M$ ,

$$\inf_{z \in M} \|x + y + z\| \le \inf_{z \in M} \left( \|x + w\| + \|y + w + z\| \right)$$
$$= \|x + w\| + \inf_{z \in M} \|y + w + z\|$$

Again we use the fact that for each  $w \in M$ ,

$$\inf_{z\in M}\|y+w+z\|=\inf_{z\in M}\|y+z\|$$

This implies that for each  $w \in M$ ,

$$\inf_{z\in M}\|x+y+z\|\leq \|x+w\|+\inf_{z\in M}\|y+z\|$$

Therefore, taking infimums over M with respect to w in this inequality yields

$$\begin{split} \|x+y+M\| &= \inf_{z \in M} \|x+y+z\| \\ &\leq \inf_{w \in M} \left( \|x+w\| + \inf_{z \in M} \|y+z\| \right) \\ &= \inf_{w \in M} \|x+w\| + \inf_{z \in M} \|y+z\| \\ &= \|x+M\| + \|y+M\| \end{split}$$

If  $\alpha = 0$ , then  $\alpha x = 0$ . Choosing  $z = 0 \in M$  gives  $\|\alpha x + M\| = 0 = |\alpha| \|x + M\|$ . Suppose that  $\alpha \neq 0$ . Then the map  $T: M \to M$  given by  $Tx = \alpha^{-1}x$  is a bijection and thus  $\inf_{z \in M} \|x + \alpha^{-1}z\| = \inf_{z \in M} \|x + z\|$ . Hence we have that

$$\begin{split} \|\alpha x + M\| &= \inf_{z \in M} \|\alpha x + z\| \\ &= \inf_{z \in M} |\alpha| \|x + \alpha^{-1} z\| \\ &= |\alpha| \inf_{z \in M} \|x + \alpha^{-1} z\| \\ &= |\alpha| \inf_{z \in M} \|x + z\| \\ &= |\alpha| \|x + M\| \end{split}$$

Suppose that ||x|| = 0. Choose a sequence  $(z_n)_{n \in \mathbb{N}} \subset M$  such that

$$\lim_{n \to \infty} ||x - z_n|| = \inf_{z \in M} ||x + z||$$
$$= 0$$

Then  $\lim_{n\to\infty} z_n = x$ . Since M is closed,  $x \in M$ . Hence x + M = 0 + M.

(2) Since M is a proper subspace, there exists  $v \in X$  such that  $v \notin M$ . Then  $||v+M|| \neq 0$ . Let  $\epsilon > 0$ . Then  $(1 - \epsilon)^{-1} ||v + M|| > ||v + M||$ . So there exists  $z \in M$  such that

$$0 < ||v + M|| \le ||v + z|| < (1 - \epsilon)^{-1} ||v + M||$$
Choose  $x = ||v + z||^{-1} (v + z)$ . Then  $||x|| = 1$  and
$$||x + M|| = ||v + z||^{-1} ||v + z + M||$$

$$= ||v + z||^{-1} ||v + M||$$

$$> 1 - \epsilon$$

(3) Let  $x \in X$ . Taking z = 0, we we see that  $||\pi(x)|| = ||x + M|| \le ||x + z|| = ||x||$ . So  $\pi$  is bounded and in particular,

$$\sup_{\|x\|=1} \|\pi(x)\| \le 1$$

From (2) we see that

$$\sup_{\|x\|=1} \|\pi(x)\| \ge 1$$

Hence  $\|\pi\| = 1$ .

(4) Suppose that X is complete. Let  $(x_i + M)_{i \in \mathbb{N}} \subset X/M$ . Suppose that  $\sum_{i \in \mathbb{N}} ||x_i + M|| < \infty$ . Let  $\epsilon > 0$ . Then for each  $i \in \mathbb{N}$ , there exists  $z_i \in M$  such that  $||x_i + z_i|| < ||x_i + M|| + \epsilon 2^{-i}$ . Define the sequence  $(a_i)_{i \in \mathbb{N}} \subset X$  by  $a_i = x_i + z_i$ . Then we have

$$\sum_{i \in \mathbb{N}} \|a_i\| = \sum_{i \in N} \|x_i + z_i\|$$

$$\leq \sum_{i \in N} \left( \|x_i + M\| + \epsilon 2^{-i} \right)$$

$$= \sum_{i \in \mathbb{N}} \|x_i + M\| + \epsilon$$

Since  $\epsilon > 0$  is arbitrary, it follows that

$$\sum_{i \in \mathbb{N}} \|a_i\| \le \sum_{i \in \mathbb{N}} \|x_i + M\| < \infty$$

Since X is complete,  $\sum_{i=1}^{\infty} a_i$  converges in X. Define  $(s_n)_{n\in\mathbb{N}} \subset X$  and  $s\in X$  by  $s_n = \sum_{i=1}^n a_i$  and  $s = \sum_{i=1}^{\infty} a_i$ . Since  $\lim_{n\to\infty} s_n = s$ , and  $\pi: X\to X/M$  is continuous, it follows that  $\lim_{n\to\infty} \pi(s_n) = \pi(s)$ . Since

$$\pi(s_n) = \sum_{i=1}^n a_i + M$$
$$= \sum_{i=1}^n x_i + M$$

We have that  $\sum_{i=1}^{\infty} x_i + M$  converges which implies that X/M is complete.

**Exercise 4.4.3.** Let X, Y be normed vector spaces and  $T \in L(X, Y)$ . Then

- (1)  $\ker T$  is closed
- (2) there exists a unique map  $S: X/\ker T \to T(X)$  such that  $T = S \circ \pi$ . Furthermore S is a bounded linear bijection and ||S|| = ||T||.

*Proof.* (1) Since T is continuous and  $\ker T = T^{-1}(\{0\})$ , we have that  $\ker T$  is closed.

(2) Suppose that there exists  $S_1, S_2 \in L(X/\ker T, T(X))$  such that  $T = S_1 \circ \pi$  and  $T = S_2 \circ \pi$ . Let  $x \in X$ . Then

$$S_1(x + \ker T) = S_1(\pi(x)) = T(x) = S_2(\pi(x)) = S_2(x + \ker T)$$

So  $S_1 = S_2$ . Therefore such a map is unique.

Define  $S: X/\ker T \to T(X)$  by  $S(x + \ker T) = T(x)$ . Then S is clearly a linear bijection that satisfies  $T = S \circ \pi$ . Let  $x \in X$  and  $z \in \ker T$ . Then

$$||S(x + \ker T)|| = ||T(x)||$$
  
=  $||T(x + z)||$   
 $\leq ||T||||x + z||$ 

Thus

$$||S(x + \ker T)|| \le ||T|| \inf_{z \in \ker T} ||x + z|| = ||T|| ||x + \ker T||$$

So S is bounded and  $||S|| \leq ||T||$ . This implies that

$$||T|| = ||S \circ \pi|| \le ||S|| ||\pi|| = ||S||$$

Thus ||S|| = ||T||.

**Exercise 4.4.4.** Let X, Y be normed vector spaces. Define  $\phi : L(X, Y) \times X \to Y$  by  $\phi(T, x) = Tx$ . Then  $\phi$  is continuous.

*Proof.* Let  $(T_1, x_1) \in L(X, Y) \times X$  and  $\epsilon > 0$ . Choose  $\delta = \min\{\frac{\epsilon}{2(\|x_1\| + \|T_1\| + 1)}, \frac{\sqrt{\epsilon}}{\sqrt{2}}\}$ . Let  $(t_2, x_2) \in L(X, Y) \times X$ . Suppose that

$$||(T_1, x_1) - (T_2, x_2)|| = \max\{||T_1 - T_2||, ||x_1 - x_2||\} < \delta$$

Then

$$\|\phi(T_{1}, x_{1}) - \phi(T_{2} - x_{2})\| = \|T_{1}x_{-}T_{2}x_{2}\|$$

$$= \|T_{1}x_{1} - T_{2}x_{1} + T_{2}x_{1} - T_{2}x_{2}\|$$

$$\leq \|(T_{1} - T_{2})x_{1}\| + \|T_{2}(x_{1} - x_{2})\|$$

$$\leq \|T_{1} - T_{2}\|\|x_{1}\| + \|T_{2}\|\|x_{1} - x_{2}\|$$

$$\leq \|T_{1} - T_{2}\|\|x_{1}\| + (\|T_{1} - T_{2}\| + \|T_{1}\|)\|x_{1} - x_{2}\|$$

$$< \delta\|x_{1}\| + (\delta + \|T_{1}\|)\delta$$

$$= \delta(\|T_{1}\| + \|x_{1}\|) + \delta^{2}$$

$$< \frac{\epsilon}{2} + \frac{\epsilon}{2}$$

$$= \epsilon$$

So  $\phi$  is continuous.

**Exercise 4.4.5.** Let X be a normed vector space and  $M \subset X$  a subspace. Then  $\overline{M}$  is a subspace.

Proof. Let  $x, y \in \overline{M}$  and  $\alpha \in \mathbb{C}$ . Then there exist sequences  $(x_n)_{n \in \mathbb{N}} \subset M$  and  $(y_n)_{n \in \mathbb{N}} \subset M$  such that  $x_n \to x$  and  $y_n \to y$ . Since M is a subspace,  $(x_n + y_n)_{n \in \mathbb{N}} \subset M$  and  $(\alpha x_n)_{n \in \mathbb{N}} \subset M$ . Since addition and scalar multiplication are continuous, we have that  $x_n + y_n \to x + y$  and  $\alpha x_n \to \alpha x$ . Thus  $x + y \in \overline{M}$  and  $\alpha x \in \overline{M}$  and hence  $\overline{M}$  is a subspace.

#### 4.5. The Hahn-Banach Theorem.

**Definition 4.5.1.** Let X be a vector space over  $\mathbb{C}$  and  $T: X \to \mathbb{C}$ . Then T is said to be a **linear functional on** X if T is linear. We define the **algebraic dual space of** X, denoted  $X^*$ , by  $X^* = \{T: X \to \mathbb{C}: T \text{ is linear}\}$ 

**Note 4.5.2.** We define  $X^*$  similarly when X is a vector space over  $\mathbb{R}$ .

**Definition 4.5.3.** Let X be a normed vector space over  $\mathbb{C}$ , and  $T: X \to \mathbb{C}$ . Then T is said to be a **bounded linear functional on** X if  $T \in L(X,\mathbb{C})$ . We define the **dual space of** X, denoted  $X^*$ , by  $X^* = L(X,\mathbb{C})$ .

**Note 4.5.4.** We define  $X^*$  similarly when X is a normed vector space over  $\mathbb{R}$ .

**Definition 4.5.5.** Let X be a vector space and  $p: X \to \mathbb{R}$ . Then p is said to be a **sublinear functional** if for each  $x, y \in X$ ,  $\lambda \ge 0$ ,

- $(1) p(x+y) \le p(x) + p(y)$
- (2)  $p(\lambda x) = \lambda p(x)$

**Exercise 4.5.6.** Let X be a vector space and  $p: X \to \mathbb{R}$  be a sublinear functional. Then p(0) = 0.

*Proof.* Set  $\lambda = 0$ . Then

$$0 = \lambda p(0)$$
$$= p(\lambda 0)$$
$$= p(0)$$

Proof. Clear

**Definition 4.5.7.** Let X be a vector space and  $p: X \to \mathbb{R}$ . Then p is said to be a **seminorm** if for each  $x, y \in X$ ,  $\lambda \in \mathbb{R}$ ,

- $(1) p(x+y) \le p(x) + p(y)$
- (2)  $p(\lambda x) = |\lambda| p(x)$

**Exercise 4.5.8.** Let X be a vector space and  $p: X \to \mathbb{R}$  be a seminorm, then p is a sublinear functional.

Proof. Clear  $\Box$ 

**Exercise 4.5.9.** Let X be a vector space and  $p: X \to \mathbb{R}$  be a seminorm. Then  $p \geq 0$ .

*Proof.* Let  $x \in X$ . Then

$$0 = p(0)$$

$$= p(x - x)$$

$$\leq p(x) + p(-x)$$

$$= p(x) + p(x)$$

$$= 2p(x)$$

So  $p(x) \geq 0$ .

**Exercise 4.5.10.** Let X be a vector space and  $p: X \to \mathbb{R}$  a sublinear functional. Then for each  $x, y \in X$ 

 $(1) -p(-x) \le p(x)$ 

(2) 
$$-p(y-x) \le p(x) - p(y) \le p(x-y)$$

Proof. Let  $x, y \in X$ .

(1) We have

$$0 = p(0)$$

$$= p(x - x)$$

$$< p(x) + p(-x)$$

So  $-p(-x) \le p(x)$ .

(2) We have

$$p(x) = p(x - y + y)$$
  

$$\leq p(x - y) + p(y)$$

So  $p(x) - p(y) \le p(x - y)$ . Switching x and y gives us  $p(y) - p(x) \le p(y - x)$  and multiplying both sides by -1 yields  $-p(y - x) \le p(x) - p(y)$ Putting these two together, we see that

$$-p(y-x) \le p(x) - p(y) \le p(x-y)$$

**Definition 4.5.11.** Let X be a normed vector space and  $p: X \to \mathbb{R}$  a sublinear functional. Then p is said to be **bounded** if there exists M > 0 such that for each  $x \in X$ ,  $p(x) \le M||x||$ .

**Exercise 4.5.12.** Let X be a normed vector space and  $p: X \to \mathbb{R}$  a sublinear functional. Then p is bounded iff p is Lipschitz.

*Proof.* Suppose that p is bounded. Then there exists M>0 such that for each  $x\in X$ ,  $p(x)\leq M\|x\|$ . Let  $x,y\in X$ . Then the previous exercise implies that

$$-M||x - y|| = -M||y - x||$$

$$\leq -p(y - x)$$

$$\leq p(x) - p(y)$$

$$\leq p(x - y)$$

$$\leq M||x - y||$$

So that

$$|p(x) - p(y)| \le M||x - y||$$

and p is Lipschitz. Conversely, suppose that p is Lipschitz. Then there exists M>0 such that for each  $x,y\in X, |p(x)-p(y)|\leq M\|x-y\|$ . Let  $x\in X$ . Then

$$p(x) \le |p(x)|$$

$$= |p(x) - p(0)|$$

$$\le M||x - 0||$$

$$\le M||x||$$

So p is bounded.

#### Theorem 4.5.13. Hahn-Banach Theorem

Let X be a vector space,  $p: X \to \mathbb{R}$  a sublinear functional,  $M \subset X$  a subspace and  $f: M \to \mathbb{R}$  a linear functional. If for each  $x \in M$ ,  $f(x) \leq p(x)$ , then there exists a linear functional  $F: X \to \mathbb{R}$  such that for each  $x \in X$ ,  $F(x) \leq p(x)$  and  $F|_M = f$ .

**Exercise 4.5.14.** Let X be a vector space and  $p: X \to \mathbb{R}$  a sublinear functional. Then there exists a linear functional  $F: X \to \mathbb{R}$  such that for each  $x \in X$ ,  $F(x) \leq p(x)$ .

*Proof.* Take  $M = \{0\}$  and  $f \equiv 0$  and apply the Hahn-Banach theorem.

Exercise 4.5.15. Equivalency of linearity (General Case) Let X be a vector space and  $p: X \to \mathbb{R}$  a sublinear functional. Then the following are equivalent:

- (1) there exists a unique  $F \in X^*$  such that  $F \leq p$
- (2) for each  $x \in X$ , -p(-x) = p(x)
- (3) p is linear

**Hint:** If there exists  $x \in X$  such that  $-p(-x) \neq p(x)$ , define  $f_1, f_2 : \operatorname{span}(x) \to \mathbb{R}$  by  $f_1(tx) = tp(x)$  and  $f_2(tx) = -tp(-x)$ 

Proof.

 $\bullet$  (1)  $\Longrightarrow$  (2):

Suppose that there exists a unique  $F \in X^*$  such that  $F \leq p$ . For the sake of contradiction, suppose that there exists  $x \in X$  such that  $-p(-x) \neq p(x)$ . Define  $f_1, f_2 : \operatorname{span}(x) \to \mathbb{R}$  by

$$f_1(tx) = tp(x)$$

and

$$f_2(tx) = -tp(-x)$$

Let  $y \in \text{span}(x)$ . Then there exists  $t \in \mathbb{R}$  such that y = tx. Then for each  $k \in \mathbb{R}$ ,

$$f_1(ky) = f_1(ktx)$$

$$= ktp(x)$$

$$= kf_1(tx)$$

$$= kf_1(y)$$

Similarly,  $f_2(ky) = kf_2(y)$  and so  $f_1, f_2 \in \text{span}(x)^*$ . If  $t \ge 0$ , then

$$f_1(y) = f_1(tx)$$

$$= tp(x)$$

$$= p(tx)$$

$$= p(y)$$

If t < 0, then

$$f_1(y) = f_1(tx)$$

$$= tp(x)$$

$$= -|t|p(x)$$

$$= -p(|t|x)$$

$$= -p(-tx)$$

$$\leq p(tx)$$

$$= p(y)$$

So  $f_1 \leq p$  on span(x). Similarly,  $f_2 \leq p$  on span(x). The Hahn-Banach theorem implies that there exist  $F_1, F_2 \in X^*$  such that  $F_1, F_2 \leq p$  and  $F_1 = f_1, F_2 = f_2$  on span(x). By the assumption of uniqueness,  $F_1 = F_2$ . This is a contradiction since

$$F_1(x) = p(x)$$

$$\neq -p(-x)$$

$$= F_2(x)$$

So for each  $x \in X$ , -p(-x) = p(x).

•  $(2) \Rightarrow (3)$ :

Suppose that for each  $x \in X$ , -p(-x) = p(x). The previous exercise implies that there exists  $F \in X^*$  such that  $F \leq p$ . Let  $x \in X$ . Then

$$-F(x) = F(-x)$$

$$\leq p(-x)$$

$$= -p(x)$$

So  $p(x) \leq F(x)$  and  $p \leq F$ . Therefore p = F and p is linear.

 $\bullet$  (3)  $\Longrightarrow$  (1):

Suppose that p is linear. Let  $F \in X^*$ . Suppose that  $F \leq p$ . Let  $x \in X$ . Then as in the case for  $(2) \implies (3)$ , we have that

$$-F(x) = F(-x)$$

$$\leq p(-x)$$

$$= -p(x)$$

which implies that p = F. So p is the unique linear function  $F \in X^*$  such that  $F \leq p$ .

**Exercise 4.5.16.** Let X be a normed vector space,  $p: X \to \mathbb{R}$  a bounded sublinear functional and  $\phi: X \to \mathbb{R}$  a linear functional. If  $\phi \leq p$ , then  $\phi \in X^*$ .

*Proof.* Since p is Lipschitz, there exists M > 0 such that for each  $x \in X$ ,

$$p(x) \le |p(x)|$$

$$\le M||x||$$

Let  $x \in X$ . Then

$$\phi(x) \le p(x)$$

$$\le |p(x)|$$

$$\le M||x||$$

and therefore

$$-M||x|| = -M||-x||$$

$$\leq -p(-x)$$

$$\leq -\phi(-x)$$

$$= \phi(x)$$

So that  $|\phi(x)| \leq M||x||$  and  $\phi \in X^*$ .

**Exercise 4.5.17.** Let X be a normed vector space and  $p: X \to \mathbb{R}$  a bounded sublinear functional. Then there exists  $\phi \in X^*$  such that for each  $x \in X$ ,  $\phi(x) \leq p(x)$ .

*Proof.* A previous exercise implies there exists  $\phi: X \to \mathbb{R}$  such that  $\phi$  is linear and  $\phi \leq p$ . The previous exercise implies that  $\phi \in X^*$ .

## Exercise 4.5.18. Equivalency of linearity (Bounded Case)

Let X be a normed vector space and  $p: X \to \mathbb{R}$  a bounded sublinear functional. Then the following are equivalent:

- (1) there exists a unique  $\phi \in X^*$  such that  $\phi \leq p$
- (2) for each  $x \in X$ , -p(-x) = p(x)
- (3) p is linear

*Proof.* Basically the same as last time.

#### Theorem 4.5.19. Complex Hahn-Banach Theorem

Let X be a vector space,  $p: X \to \mathbb{R}$  a seminorm,  $M \subset X$  a subspace and  $f: M \to \mathbb{C}$  a linear functional. If for each  $x \in M$ ,  $|f(x)| \leq p(x)$ , then there exists a linear functional  $F: X \to \mathbb{C}$  such that for each  $x \in X$ ,  $|F(x)| \leq p(x)$  and  $F|_M = f$ .

**Exercise 4.5.20.** Let X be a normed vector space,  $M \subset X$  a subspace and  $f \in M^*$ . Then there exists  $F \in X^*$  such that ||F|| = ||f|| and  $F|_M = f$ .

Proof. If f = 0, Choose F = 0. Suppose  $f \neq 0$ . Then  $||f|| \neq 0$  and there exists  $x_0 \in M$  such that  $x_0 \neq 0$ . Thus  $||f|| = \sup\{|f(x)| : x \in M \text{ and } ||x|| = 1\}$ . Define  $p : X \to [0, \infty)$  by p(x) = ||f|| ||x||. Then p is a sublinear functional on X and for each  $x \in M$ ,  $|f(x)| \leq p(x)$ . So there exists a linear functional  $F : X \to \mathbb{C}$  such that for each  $x \in X$ ,  $|F(x)| \leq p(x) = ||f|| ||x||$  and  $F|_M = f$ . Thus  $F \in X^*$  with  $||F|| \leq ||f||$ . Also

$$||F|| = \sup_{\substack{x \in X \\ ||x|| = 1}} |F(x)| \ge \sup_{\substack{x \in M \\ ||x|| = 1}} |F(x)| = \sup_{\substack{x \in M \\ ||x|| = 1}} |f(x)| = ||f||$$

So ||F|| = ||f||.

**Exercise 4.5.21.** Let X be a normed vector space,  $M \subsetneq X$  a proper closed subspace and  $x \in X \setminus M$ . Then there exists  $F \in X^*$  such that  $F|_M = 0$ , ||F|| = 1 and  $F(x) = ||x+M|| \neq 0$ . **Hint:** Consider  $f: M + \mathbb{C}x \to \mathbb{C}$  defined by  $f(m + \lambda x) = \lambda ||x + M||$ .

*Proof.* Define  $f: M + \mathbb{C}x \to \mathbb{C}$  as above. Clearly f is linear and  $f|_M = 0$ . Let  $m \in M$  and  $\lambda \in \mathbb{C}$ . If  $\lambda = 0$ , then  $|f(m + \lambda x)| = 0 \le ||m + \lambda x||$ . Suppose that  $\lambda \ne 0$ . Then

$$|f(m + \lambda x)| = |\lambda| ||x + M||$$

$$= ||\lambda x + M||$$

$$= \inf_{z \in M} ||z + \lambda x||$$

$$\leq ||m + \lambda x||$$

So  $f \in (M + \mathbb{C}x)^*$  and  $||f|| \le 1$ . Let  $\epsilon > 0$ . A previous exercise tells us that there exist  $m \in M, \lambda \in \mathbb{C}$  such that  $||m + \lambda x|| = 1$  and  $||m + \lambda x + M|| > 1 - \epsilon$ . Then

$$|f(m + \lambda x)| = |\lambda| ||x + M||$$

$$= ||\lambda x + M||$$

$$= ||m + \lambda x + M||$$

$$> 1 - \epsilon$$

So

$$||f|| = \sup_{\substack{z \in M + \mathbb{C}x \\ ||z|| = 1}} |f(z)| \ge 1$$

Hence ||f|| = 1. The same exercise also tells us that  $f(x) = ||x+M|| \neq 0$ . Using the previous exercise, there exists  $F \in X^*$  such that ||F|| = ||f|| = 1 and  $F|_{M+\mathbb{C}x} = f$ .

**Exercise 4.5.22.** Let X be a normed vector space and  $x \in X$ . If  $x \neq 0$ , then there exists  $F \in X^*$  such that ||F|| = 1 and F(x) = ||x||.

*Proof.* Define  $f: \mathbb{C}x \to \mathbb{C}$  by  $f(\lambda x) = \lambda ||x||$ . Then f is linear and f(x) = ||x||. Clearly

$$\sup_{\substack{z \in \mathbb{C}x \\ \|z\| = 1}} |f(z)| = 1$$

So  $f \in (\mathbb{C}x)^*$  and ||f|| = 1. By a previous exercise, there exists  $F \in X^*$  such that ||F|| = ||f|| = 1 and  $F|_{\mathbb{C}x} = f$ .

**Exercise 4.5.23.** Let X be a normed vector space and  $x \in X$ . Then x = 0 iff for each  $\phi \in X^*$ ,  $\phi(x) = 0$ .

*Proof.* Clear by previous exercise.

**Exercise 4.5.24.** Let X be a normed vector space. Then  $X^*$  separates the points of X.

*Proof.* Let  $x, y \in X$ . Suppose that  $x \neq y$ . Then  $x - y \neq 0$ . The previous exercies implies that there exists  $F \in X^*$  such that ||F|| = 1 and

$$F(x) - F(y) = F(x - y) = ||x - y|| \neq 0$$

Thus  $F(x) \neq F(y)$  and  $X^*$  separates the points of X.

**Exercise 4.5.25.** Let X be a normed vector space and  $x \in X$ . Define  $\hat{x}: X^* \to \mathbb{C}$  by  $\hat{x}(f) = f(x)$ . Then  $\hat{x} \in X^{**}$  and  $\|\hat{x}\| = \|x\|$ .

*Proof.* Let  $f, g \in X^*$  and  $\lambda \in \mathbb{C}$ . Then

$$\hat{x}(f + \lambda g) = (f + \lambda g)(x) = f(x) + \lambda g(x) = \hat{x}(f) + \lambda \hat{x}(g)$$

So  $\hat{x}$  is linear. For each  $f \in X^*$ ,

$$|\hat{x}(f)| = |f(x)| \le ||x|| ||f||$$

Hence  $\hat{x} \in X^{**}$  with  $\|\hat{x}\| \leq \|x\|$ . If x = 0, then  $\hat{x} = 0$  and  $\|\hat{x}\| = \|x\|$ . Suppose that  $x \neq 0$ . Then a previous exercise implies that there exists  $F \in X^*$  such that  $\|F\| = 1$  and  $F(x) = \|x\|$ . Then we have that

$$\sup_{\substack{f \in X^* \\ \|f\|=1}} |\hat{x}(f)| = \sup_{\substack{f \in X^* \\ \|f\|=1}} |f(x)| \ge |F(x)| = \|x\|$$

Hence  $||\hat{x}|| = ||x||$ .

**Exercise 4.5.26.** Let X be a normed vector space. Define  $\phi: X \to X^{**}$  by  $\phi(x) = \hat{x}$ . Then  $\phi$  is a linear isometry.

*Proof.* Let  $x, y \in X$  and  $\lambda \in \mathbb{C}$ . Then for each  $f \in X^*$ , we have that

$$\phi(x + \lambda y)(f) = \widehat{x + \lambda y}(f)$$

$$= f(x + \lambda y)$$

$$= f(x) + \lambda f(y)$$

$$= \widehat{x}(f) + \lambda \widehat{y}(f)$$

$$= \phi(x)(f) + \lambda \phi(y)(f)$$

So  $\phi(x+\lambda y)=\phi(x)+\lambda\phi(y)$  and  $\phi$  is linear. The previous exercise tells us that

$$\|\phi(x) - \phi(y)\| = \|\phi(x - y)\|$$

$$= \|\widehat{x - y}\| = \|x - y\|$$

So  $\phi$  is an isometry.

**Definition 4.5.27.** Let X be a normed vector space and define  $\phi: X \to X^{**}$  as above. We define  $\widehat{X} = \phi(X) \subset X^{**}$ . Since  $\widehat{X}$  and X are isomorphic, we may identify X as a subset of  $X^{**}$ .

**Definition 4.5.28.** Let X be a normed vector space and define  $\phi: X \to X^{**}$  as above. Then X is said to be reflexive if  $\phi$  is surjective. In this case  $\phi$  is then an isomorphism

**Exercise 4.5.29.** Let X be a normed vector space and  $f: X \to \mathbb{C}$  a linear functional on X. Then f is bounded iff ker f is closed.

*Proof.* Suppose that f is continuous. Since  $\{0\}$  is closed, we have that  $\ker f = f^{-1}(\{0\})$  is closed. Conversely, suppose that  $\ker f$  is closed. If  $\ker f = X$ , then f = 0 and f is continuous. Suppose that  $\ker f \neq X$ . Then  $\ker f$  is a proper, closed subspace of X. A previous exercise tells us that there exists  $x \in X$  such that ||x|| = 1 and  $||x + \ker f|| > \frac{1}{2}$ .

Let  $y \in X$ . Suppose that  $||y|| < \frac{1}{2}$ . Then for each  $z \in \ker f$ ,

$$||z - (x + y)|| = ||(z - x) - y||$$

$$\ge ||z - x|| - ||y||$$

$$> \frac{1}{2} - \frac{1}{2}$$

$$= 0$$

So  $x + y \notin \ker f$ . Therefore  $f(B(x, \frac{1}{2})) \cap \{0\} = \emptyset$ . If  $f(B(x, \frac{1}{2}))$  is unbounded, then  $f(B(x, \frac{1}{2})) = \mathbb{C}$  by linearity. This is a contradiction since  $0 \notin f(B(x, \frac{1}{2}))$ . So There exists s > 0 such that  $f(B(x, \frac{1}{2})) \subset B(0, s)$  and thus f is bounded.

#### **Exercise 4.5.30.** Let X be a normed vector space.

- (1) Let  $M \subseteq X$  be a proper closed subspace of X and  $x \in X \setminus M$ . Then  $M + \mathbb{C}x$  is closed.
- (2) Let  $M \subset X$  be a finite dimensional subspace of X. Then M is closed.
- Proof. (1) Let  $y \in X$  and  $(y_n)_{n \in \mathbb{N}} \subset M + \mathbb{C}x$ . Suppose that  $y_n \to y$ . If  $y \in M$ , then  $y \in M + \mathbb{C}x$ . Suppose that  $y \notin M$ . For each  $n \in \mathbb{N}$ , there exists  $m_n \in M$  and  $\lambda_n \in \mathbb{C}$  such that  $y_n = m_n + \lambda_n x$ . A previous exercise tells us that there exists  $F \in X^*$  such that ||F|| = 1,  $|F||_M = 0$  and  $|F|(x) = ||x + M|| \neq 0$ . Since  $|F|(x) \to F(y)$  is continuous,  $|F|(y) \to F(y)$ . Since for each  $|F|(x) \to F(y)$  is continuous,

$$F(y_n) = F(m_n + \lambda_n x) = F(m_n) + \lambda_n(F_x) = \lambda_n F(x)$$

we have that  $\lambda_n F(x) \to F(y)$ . Since  $F(x) \neq 0$ , this implies that  $\lambda_n \to F(x)^{-1} F(y)$ . It follows that  $\lambda_n x \to F(x)^{-1} F(y) x$ . Since for each  $n \in \mathbb{N}$ ,  $m_n = y_n - \lambda_n x$ , we know that  $m_n \to y - F(x)^{-1} F(y) x$ . Since  $(m_n)_{n \in \mathbb{N}} \subset M$  and M is closed, we have that  $y - F(x)^{-1} F(y) x \in M$  and therefore  $y \in M + \mathbb{C}x$ . Hence  $M + \mathbb{C}x$  is closed.

(2) If M = X, then M is closed. Suppose that  $M \neq X$ . Let  $(x_i)_{i=1}^n$  be a basis for M. Define  $N_0 = \{0\}$  and for each  $i = 1, 2, \dots, n$ , define  $N_i = N_{i-1} + \mathbb{C}x_i$ . Since  $N_0$  is a proper closed subpace of X and  $x_1 \in X \setminus N_0$ , (1) implies that  $N_1$  is closed. Proceed inductively to obtain that  $M = N_n$  is closed.

**Exercise 4.5.31.** Let X be an infinite-dimensional normed vector space.

(1) There exists a sequence  $(x_n)_{n\in\mathbb{N}}\subset X$  such that for each  $m,n\in\mathbb{N}, \|x_n\|=1$  and if  $m\neq n$ , then  $\|x_m-x_n\|>\frac{1}{2}$ .

- (2) X is not locally compact.
- Proof. (1) Define  $N_0 = \{0\}$ . Then  $N_0$  is a closed proper subspace of X. Choose  $x_1 \in X$  such that  $||x_1|| = 1$ . Using the results of previous exercises, we proceed inductively. For each  $n \geq 2$  we define  $N_{n-1} = \operatorname{span}(x_1, x_2, \cdots, x_{n-1})$ . Then  $N_{n-1}$  is a closed proper subspace of X. Thus we may choose  $x_n \in X$  such that  $||x_n|| = 1$  and  $||x_n + N_{n-1}|| > \frac{1}{2}$ . Let  $m, n \in \mathbb{N}$ . Suppose that m < n. Then  $x_m \in N_{n-1}$ . Thus  $||x_n x_m|| \geq ||x_n + N_{n-1}|| > \frac{1}{2}$

(2) Suppose that X is locally compact. Then  $\overline{B(0,1)}$  is compact and therefore sequentially compact. Using  $(x_n)_{n\in\mathbb{N}}\subset \overline{B(0,1)}$  defined in (1), we see that there exists a subsequence  $(x_{n_k})_{k\in\mathbb{N}}$ ,  $x\in \overline{B(0,1)}$  such that  $x_{n_k}\to x$ . Then  $(x_{n_k})_{k\in\mathbb{N}}$  is Cauchy. So there exists  $N\in N$  such that for each  $j,k\in\mathbb{N}$ , if  $j,k\geq N$ , then  $||x_{n_j}-x_{n_k}||<\frac{1}{2}$ . Then  $||x_{n_N}-x_{n_{N+1}}||<\frac{1}{2}$ . This is a contradiction since by construction,  $||x_{n_N}-x_{n_{N+1}}||>\frac{1}{2}$ . Thus X is not locally compact.

**Exercise 4.5.32.** Let X, Y be normed vector spaces and  $T \in L(X, Y)$ .

- (1) Define the **adjoint of** T, denoted  $T^*: Y^* \to X^*$  by  $T^*(f) = f \circ T$ . Then  $T^* \in L(Y^*, X^*)$ .
- (2) Applying the result from (1) twice, we have that  $T^{**} \in L(X^{**}, Y^{**})$ . We have that for each  $x \in X$ ,  $T^{**}(\hat{x}) = \widehat{T(x)}$ .
- (3)  $T^*$  is injective iff T(X) is dense in Y.
- (4) If  $T^*(Y^*)$  is dense in  $X^*$ , then T is injective. The converse is true if X is reflexive.

*Proof.* (1) Let  $f \in Y^*$ . Then  $||T^*(f)|| = ||f \circ T|| \le ||T|| ||f||$ . So  $T^* \in L(Y^*, X^*)$  with  $||T^*|| \le ||T||$ .

(2) Let  $x \in X$ . Let  $f \in Y^*$ . Then

$$T^{**}(\hat{x})(f) = \hat{x} \circ T^{*}(f)$$

$$= \hat{x}(T^{*}(f))$$

$$= \hat{x}(f \circ T)$$

$$= f \circ T(x)$$

$$= f(T(x))$$

$$= \widehat{T(x)}(f)$$

Hence  $T^{**}(\hat{x}) = \widehat{T(x)}$ .

(3) Suppose that T(X) is not dense in Y. Then  $\overline{T(X)} \neq \underline{Y}$ . So T(X) is a proper closed subspace of Y and there exists  $y \in Y$  such that  $y \notin \overline{T(X)}$ . By a previous exercise, there exists  $f \in Y^*$  such that  $f(y) = \|y + \overline{T(X)}\| \neq 0$ ,  $\|f\| = 1$  and  $f|_{\overline{T(X)}} = 0$ . Let  $x \in X$ . Then  $T^*(f)(x) = f \circ T(x) = 0$ . Hence  $T^*(f) = 0 = T^*(0)$ . Since  $f \neq 0$ ,  $T^*$  is not injective. Now suppose that T(X) is dense in Y. Let  $f, g \in Y^*$ . Define  $h \in Y^*$  by h = f - g

Now suppose that T(X) is dense in Y. Let  $f, g \in Y^*$ . Define  $h \in Y^*$  by h = f - g. Suppose that  $T * (f) = T^*(g)$ . Then  $T^*(h) = 0$ . So for each  $x \in X$ , h(T(x)) = 0. Let  $y \in Y$  and  $\epsilon > 0$ . By continuity, there exists  $\delta > 0$  such that for each  $y' \in Y$ , if  $||y - y'|| < \delta$ , then  $||h(y) - h(y')|| < \epsilon$ . Since T(X) is dense in Y, there exists  $x \in X$ 

such that  $||y - T(x)|| < \delta$ . Thus

$$||h(y)|| \le ||h(y) - h(T(x))|| + ||h(T(x))||$$
  
=  $||h(y) - h(T(x))||$   
<  $\epsilon$ 

Since  $\epsilon > 0$  is arbitrary, ||h(y)|| = 0. This implies that h(y) = 0 and therefore f(y) = g(y). Since  $y \in Y$  is arbitrary, f = g and  $T^*$  is injective.

(4) For the sake of contradiction, suppose that  $T^*(Y^*)$  is dense in  $X^*$  and T is not injective. Then there exist  $x_1, x_2 \in X$  such that  $x_1 \neq x_2$  and  $T(x_1) = T(x_2)$ . Define  $x = x_1 - x_2$ . Then  $x \neq 0$  and T(x) = 0. A previous exercise implies that there exists  $F \in X^*$  such that  $F(x) = ||x|| \neq 0$  and ||F|| = 1. Let  $\epsilon > 0$ . Choose  $g \in Y^*$  such that  $||F - T^*(g)|| < \epsilon$ . Then

$$||x|| = |F(x)|$$

$$\leq |F(x) - T^*(g)(x)| + |T^*(g)(x)|$$

$$< \epsilon ||x|| + |g(T(x))|$$

$$= \epsilon ||x||$$

Since  $\epsilon > 0$  is arbitrary, we have that ||x|| = 0 which is a contradiction. Hence if  $T^*(Y^*)$  is dense in  $X^*$ , then T is injective.

Now, suppose that X is reflexive and T is injective. Let  $\phi_1, \phi_2 \in X^{**}$ . Suppose that  $T^{**}(\phi_1) = T^{**}(\phi_2)$ . Then  $T^{**}(\phi_1 - \phi_2) = 0$ . Since X is reflexive, there exist  $x_1, x_2 \in X$  such that  $\phi_1 = \hat{x_1}$  and  $\phi_2 = \hat{x_2}$ . Define  $x = x_1 - x_2$ . Then  $T^{**}(\hat{x}) = 0$ . So for each  $f \in Y^*$ ,

$$T^{**}(\hat{x})(f) = \hat{x} \circ T^{*}(f)$$

$$= \hat{x}(T^{*}(f))$$

$$= \hat{x}(f \circ T)$$

$$= f \circ T(x)$$

$$= f(T(x))$$

$$= 0$$

Suppose that  $T(x) \neq 0$ . Then a previous exercise implies that there exists  $g \in Y^*$  such that  $g(T(x)) = ||T(x)|| \neq 0$  and ||g|| = 1. This is a contradiction since g(T(x)) = 0. So T(x) = 0. Since T is injective, this implies that x = 0. Hence  $\hat{x} = 0$  and thus  $\phi_1 = \phi_2$ . Thus  $T^{**}$  is injective. By (3), we have that  $T^*(Y^*)$  is dense in  $X^*$ .

**Exercise 4.5.33.** Let X be a normed vector space. Then X is reflexive iff  $X^*$  is reflexive.

*Proof.* Suppose that X is reflexive. Let  $\alpha \in X^{***}$ . Define  $f: X \to \mathbb{C}$  by  $f(x) = \alpha(\hat{x})$ . Clearly f is linear and a previous exercise tells us that for each  $x \in X$ ,

$$|f(x)| \le ||\alpha|| ||\hat{x}||$$
$$= ||\alpha|| ||x||$$

So  $f \in X^*$ . Let  $\phi \in X^{**}$ . Since X is reflexive, there exists  $x \in X$  such that  $\phi = \hat{x}$ . Then

$$\alpha(\phi) = \alpha(\hat{x})$$

$$= f(x)$$

$$= \hat{x}(f)$$

$$= \hat{f}(\hat{x})$$

$$= \hat{f}(\phi)$$

Hence  $\alpha = \hat{f}$ . Thus the map  $X^* \to X^{***}$  given by  $f \mapsto \hat{f}$  is surjective and so  $X^*$  is reflexive.

Conversely, suppose that  $X^*$  is reflexive. Since  $\phi: X \to X^{**}$  given by  $\phi(x) = \hat{x}$  is an isometry,  $\widehat{X} \subset X^{**}$  is closed. For the sake of contradiction, suppose that  $\widehat{X} \neq X^{**}$ . Then there exists  $\alpha \in X^{**}$  such that  $\alpha \notin \widehat{X}$ . Thus there exists  $F \in X^{***}$  such that  $\|F\| = 1$ ,  $F(\alpha) = \|\alpha + \widehat{X}\| \neq 0$  and  $F|_{\widehat{X}} = 0$ . Since  $X^*$  is reflexive, there exists  $f \in X^*$  such that  $F = \widehat{f}$ . A previous exercise tells us that  $\|f\| = \|\widehat{f}\| = \|F\| = 1$ . Since for each  $x \in X$ ,  $f(x) = \widehat{x}(f) = \widehat{f}(\widehat{x}) = F(\widehat{x}) = 0$ , we have that f = 0. Thus  $\|f\| = 0$ , a contradiction. So  $\widehat{X} = X^{**}$  and X is reflexive.

## 4.6. Multilinear Maps.

**Definition 4.6.1.** Let  $X_1, \dots, X_n, Y$  be normed vector spaces and  $T : \prod_{i=1}^n X_i \to Y$  multilinear. Then T is said to be **bounded** if there exists  $C \ge 0$  such that for each  $x_1, \dots, x_n \in X$ ,

$$||T(x_1, \cdots, x_n)|| \le C||x_1|| \cdots ||x_n||$$

We define

$$L^n(X_1,\ldots,X_n;Y) = \left\{T: \prod_{i=1}^n X_i \to Y: T \text{ is multilinear and bounded}\right\}$$

If  $X_1 = \cdots = X_n = X$ , we write  $L^n(X, Y)$  in place of  $L^n(X, \ldots, X; Y)$ . If  $X_1 = \cdots = X_n = Y = X$ , we write  $L^n(X)$ .

**Note 4.6.2.** For the remainder of this section we will primarily consider  $L^2(X_1, X_2; Y)$  to avoid notational clutter, but all results immediately generalize to  $L^n(X_1, \ldots, X_n; Y)$ 

**Exercise 4.6.3.** Let  $X_1, X_2$  and Y be normed vector spaces and  $T: X_1 \times X_2 \to Y$  bilinear. Then the following are equivalent:

- (1) T is continuous
- (2) T is continuous at (0,0)
- (3) T is bounded

Proof.

- $(1) \Longrightarrow (2)$ : Trivial
- $\bullet$  (2)  $\Longrightarrow$  (3):

Suppose that T is continuous at (0,0). For the sake of contradiction, suppose that T is not bounded. Then for each  $C \geq 0$ , there exist  $(x_1, x_2) \in X_1 \times X_2$  such that  $||T(x_1, x_2)|| \geq C||x_1|| ||x_2||$ . Hence there exist  $(a_n)_{n \in \mathbb{N}} \subset X_1$  and  $(b_n)_{n \in \mathbb{N}} \subset X_2$  such that for each  $n \in \mathbb{N}$ ,  $||T(a_n, b_n)|| \geq n^2 ||a_n|| ||b_n||$ . Hence for each  $n \in \mathbb{N}$ ,  $||a_n||$ ,  $||b_n|| geq 0$ . Define

$$(a'_n)_{n\in\mathbb{N}}\subset X_1$$

and  $(b'_n)_{n\in\mathbb{N}}\subset X_2$  by  $a'_n=\frac{a_n}{n\|a_n\|}$  and  $b'_n=\frac{b_n}{n\|b_n\|}$ . Then  $(a'_n,b'_n)\to (0,0)$ . Continuity implies that  $T(a'_n,b'_n)\to 0$ . By construction, for each  $n\in\mathbb{N}$ ,

$$||T(a'_n, b'_n)|| = \frac{1}{n^2 ||a_n|| ||b_n||} T(a_n, b_n)$$

$$\geq \frac{n^2 ||a_n|| ||b_n||}{n^2 ||a_n|| ||b_n||}$$

$$= 1$$

which is a contradiction. So T is bounded.

• (3)  $\Longrightarrow$  (1): Suppose that T is bounded. Then there exists C > 0 such that for each  $(x_1, x_2) \in X_1 \times X_2$ ,  $||T(x_1, x_2)|| \le C||x_1|| ||x_2||$ . Let  $(a, b) \in X_1 \times X_2$  and  $(a_n, b_n)_{n \in \mathbb{N}} \subset X_1 \times X_2$ . Suppose that  $(a_n, b_n) \to (a, b)$ . Then  $a_n \to a$ ,  $b_n \to b$  and  $(a_n)_{n \in \mathbb{N}}$ ,  $(b_n)_{n \in \mathbb{N}}$  are bounded. So there exists  $B \geq 0$  such that for each  $n \in \mathbb{N}$   $||b_n|| \leq B$ . Hence

$$||T(a_n, b_n) - T(a, b)|| = ||T(a_n, b_n) - T(a, b_n) + T(a, b_n) - T(a, b)||$$

$$\leq ||T(a_n, b_n) - T(a, b_n)|| + ||T(a, b_n) - T(a, b)||$$

$$= ||T(a_n - a, b_n)|| + ||T(a, b_n - b)||$$

$$\leq C(||a_n - a|| ||b_n|| + ||a|| ||b_n - b||)$$

$$\leq C(||a_n - a||B + ||a|| ||b_n - b||)$$

$$\to 0$$

Thus T is continuous.

**Definition 4.6.4.** Let  $X_1, X_2$  and Y be normed vector spaces and  $T \in L^2(X_1, X_2; Y)$ . We define the **operator norm** on  $L^2(X_1, X_2; Y)$ , denoted  $\|\cdot\|: L^2(X_1, X_2; Y) \to [0, \infty)$ , by

$$||T|| = \inf\{C \ge 0 : \text{for each } (x_1, x_2) \in X_1 \times X_2, ||T(x_1, x_2)|| \le C||x_1|| ||x_2||\}$$

**Exercise 4.6.5.** Let  $X_1, X_2$  and Y be normed vector spaces. If  $X_1 \neq \{0\}$  and  $X_2 \neq \{0\}$ , then the operator norm on L(X,Y) is given by:

(1) 
$$||T|| = \sup_{\|x_1\| = 1} ||T(x_1, x_2)||$$

(2) 
$$||T|| = \sup_{x_1 \neq 0} ||x_1||^{-1} ||x_2||^{-1} ||T(x_1, x_2)||$$

$$\begin{aligned} &(1) \ \|T\| = \sup_{\|x_1\|=1, \|x_2\|=1} \|T(x_1, x_2)\| \\ &(2) \ \|T\| = \sup_{x_1 \neq 0, x_2 \neq 0} \|x_1\|^{-1} \|x_2\|^{-1} \|T(x_1, x_2)\| \\ &(3) \ \|T\| = \inf\{C \geq 0 : \text{for each } (x_1, x_2) \in X_1 \times X_2, \ \|T(x_1, x_2)\| \leq C \|x_1\| \|x_2\| \} \end{aligned}$$

*Proof.* Since  $X_1 \neq \{0\}$  and  $X_2 \neq \{0\}$ , the supremums in (1) and (2) are well defined. Let  $T \in L^2(X_1, X_2; Y)$ . Bilinearity of T implies that the sets over which the supremums are taken in (1) and (2) are the same. So (1) and (2) are equal. Now, set

$$M = \sup_{\|x_1\|=1, \|x_2\|=1} \|T(x_1, x_2)\|$$

and

$$m = \inf\{C \ge 0 : \text{for each } (x_1, x_2) \in X_1 \times X_2, \|T(x_1, x_2)\| \le C\|x_1\| \|x_2\| \}$$

Let  $(x_1, x_2) \in X_1 \times X_2$ . If  $||x_1|| = 0$  or  $||x_2|| = 0$ , then  $T(x_1, x_2) = 0$  and  $||T(x_1, x_2)|| \le 1$  $M||x_1|| ||x_2||$ . Suppose that  $||x_1|| \neq 0$  and  $||x_2|| \neq 0$ . Then

$$||T(x_1, x_2)|| = \left( ||T(||x_1||^{-1}x_1, ||x_2||^{-1}x_2)|| \right) ||x_1|| ||x_2||$$

$$\leq M||x_1|| ||x_2||$$

Hence  $M \in \{C \geq 0 : \text{ for each } x \in X, \|Tx\| \leq C\|x\|\}$  and  $m \leq M.$  Let  $C \in \{C \geq 1\}$ 0: for each  $(x_1, x_2) \in X_1 \times X_2$ ,  $||T(x_1, x_2)|| \leq C||x_1|| ||x_2||$ . Suppose that  $||x_1|| = 1$  and  $||x_2|| = 1$ . Then  $||T(x_1, x_2)|| \le C||x_1|| ||x_2|| = C$ . So  $M \le C$ . Therefore  $M \le m$ . So M = mand the supremum in (1) is the same as the infimum in (3). 

**Exercise 4.6.6.** Let  $X_1, X_2$  and Y be normed vector spaces. Then  $\|\cdot\|: L^2(X_1, X_2; Y) \to \mathbb{R}$  $[0,\infty)$  is a norm.

Proof. 

**Exercise 4.6.7.** Let  $X_1, X_2, Y$  be normed vector spaces and  $T_1 \in L(X_1, L(X_2, Y))$ . Define  $T: X_1 \times X_2 \to Y$  by  $T(x_1, x_2) = T_1(x_1)(x_2)$ . Then  $T \in L^2(X_1, X_2; Y)$ .

*Proof.* It is straightforward to show that T is multilinear. For  $x_1 \in X_1$  and  $x_2 \in X_2$ ,

$$||T(x_1, x_2)|| = ||T_1(x_1)(x_2)||$$

$$\leq ||T_1(x_1)|| ||x_2||$$

$$\leq ||T_1|| ||x_1|| ||x_2||$$

So  $T \in L^2(X_1, X_2; Y)$ .

**Exercise 4.6.8.** Let  $X_1, X_2, Y$  be normed vector spaces and  $T \in L^2(X_1, X_2; Y)$ . Define the map  $T_1: X_1 \to Y^{X_2}$  by  $T_1(x_1)(\cdot) = T(x_1, \cdot)$ . Then  $T_1 \in L(X_1, L(X_2, Y))$ .

*Proof.* Let  $x_1 \in X_1$ . By definition of T,  $T_1(x_1)$  is linear. Since T is bounded, there exists  $C \geq 0$  such that for each  $a_1 \in X_1$ ,  $a_2 \in X_2$ ,  $T(a_1, a_2) \leq C ||a_1|| ||a_2||$ . Then for each  $x_2 \in X_2$ ,

$$||T_1(x_1)(x_2)|| = ||T(x_1, x_2)||$$
  

$$\leq (C||x_1||)||x_2||$$

So  $T_1(x_1) \in L(X_2, Y)$  with  $||T_1(x_1)|| \leq C||x_1||$ . Since  $x_1 \in X_1$  was arbitrary,  $T_1 : X_1 \to L(X, Y)$ . By definition of T,  $T_1$  is linear. The preceding argument tells us that for each  $x_1 \in X_1$ ,

$$||T_1(x_1)|| \le C||x_1||$$

So  $T_1 \in L(X_1, L(X_2, Y))$  with  $||T_1|| \le C$ .

**Exercise 4.6.9.** Let  $X_1, X_2$  be normed vector spaces. Define a map  $\phi : L^2(X_1, X_2; Y) \to L(X_1, L(X_2, Y))$  by  $\phi(T)(x_1)(x_2) = T(x_1, x_2)$ . Then T is an isometric isomorphism.

$$Proof.$$
 .

**Definition 4.6.10.** Let  $X_1, X_2$  be normed vector spaces,  $\phi_1 \in X_1^*$  and  $\phi_2 \in X_2^*$ . Define  $\phi_1 \otimes \phi_2 : X_1 \times X_2$  by  $\phi_1 \otimes \phi_2(x_1, x_2) = \phi_1(x_1)\phi_2(x_2)$ .

**Exercise 4.6.11.** Let  $X_1, X_2$  be normed vector spaces,  $\phi_1 \in X_1^*$  and  $\phi_2 \in X_2^*$ . Then  $\phi_1 \otimes \phi_2 \in L^2(X_1, X_2; \mathbb{C})$ .

Proof. Clear. 
$$\Box$$

**Exercise 4.6.12.** Let  $X_1, X_2$  be normed vector spaces and  $(x_1, x_2) \in X_1 \times X_2$ . If for each  $\phi_1 \in X_1^*$  and  $\phi_2 \in X_2^*$ ,  $\phi_1 \otimes \phi_2(x_1, x_2) = 0$ , then  $x_1 = 0$  or  $x_2 = 0$ .

*Proof.* Suppose that  $x_1 \neq 0$  and  $x_2 \neq 0$ . The previous section implies that there exist  $\phi_1 \in X_1^*$  and  $\phi_2 \in X_2^*$  such that  $\phi_1(x_1) = ||x_1|| \neq 0$  and  $\phi_2(x_2) = ||x_2|| \neq 0$ . Then

$$\phi_1 \otimes \phi_2(x_1, x_2) = \phi_1(x_1)\phi_2(x_2)$$

$$\neq 0$$

#### 4.7. Tensor Products.

**Note 4.7.1.** This section assumes familiarity with the algebraic tensor product of two vector spaces. See section ??? of [1] for details.

**Definition 4.7.2.** Let X, Y and Z be Banach spaces and  $\phi \in L^2(X, Y; Z)$ . Then  $(Z, \phi)$  is said to be a **tensor product** of X with Y if

- (1) span  $\phi(X \times Y)$  is dense in Z
- (2) for each Banach space W and  $\psi \in L^2(X, Y; W)$ , there exists a unique  $\psi' \in L(Z, W)$  such that  $\psi' \circ \phi = \psi$ , i.e. such that the following diagram commutes:

$$X \times Y \xrightarrow{\phi} Z$$

$$\downarrow^{\psi'}$$

$$W$$

If  $(Z, \phi)$  is a tensor product of X with Y. We often write  $Z = X \otimes Y$  and for each  $x \in X$ ,  $y \in Y$ , we often write  $\phi(x, y) = x \otimes y$ .

**Exercise 4.7.3.** Let X and Y be Banach spaces,  $U \subset X$  and  $V \subset Y$ . Set  $W = \{u \otimes v : u \in U \text{ and } v \in V\} \subset X \otimes Y$ . If U and V are linearly independent, then W is linearly independent.

**Hint:** For  $\phi \in X^*$ ,  $\psi \in Y^*$ , define  $T \in L^2(X,Y;\mathbb{C})$  by  $T(x,y) = \phi(x)\psi(y)$ .

*Proof.* Let  $w = \sum_{u \in U} \sum_{v \in V} \lambda_{u,v} u \otimes v$ . Suppose that w = 0. Let  $\phi \in X^*$  and  $\psi \in Y^*$ . Define  $T \in L^2(X,Y;\mathbb{C})$  by  $T(x,y) = \phi(x)\psi(y)$ . By definition of the tensor product, there exists a unique  $T' \in L(X \otimes Y,\mathbb{C})$  such that for each  $x \in X$  and  $y \in Y$ ,  $T'(x \otimes y) = T(x,y)$ . Then

$$0 = T'(w)$$

$$= T'(\sum_{u \in U} \sum_{v \in V} \lambda_{u,v} u \otimes v)$$

$$= \sum_{u \in U} \sum_{v \in V} \lambda_{u,v} T'(u \otimes v)$$

$$= \sum_{u \in U} \sum_{v \in V} \lambda_{u,v} T(u,v)$$

$$= \sum_{u \in U} \sum_{v \in V} \lambda_{u,v} \phi(u) \psi(v)$$

$$= \phi\left(\sum_{u \in U} \sum_{v \in V} \lambda_{u,v} \psi(v) u\right)$$

Since  $\phi \in X^*$  is arbitary, a previous exercise in the section on linear functionals implies that

$$0 = \sum_{u \in U} \sum_{v \in V} \lambda_{u,v} \psi(v) u$$
$$= \sum_{v \in U} \left( \sum_{v \in V} \lambda_{u,v} \psi(v) \right) u$$

Linear independence of U implies that for each  $u \in U$ ,

$$0 = \sum_{v \in V} \lambda_{u,v} \psi(v)$$
$$= \psi\left(\sum_{v \in V} \lambda_{u,v} v\right)$$

Since  $\psi \in Y^*$  is arbitary, for each  $u \in U$ ,

$$\sum_{v \in V} \lambda_{u,v} v = 0$$

Linear independence of V implies that for each  $u \in U, v \in V$ ,  $\lambda_{u,v} = 0$ . Hence W is linearly independent.

# Exercise 4.7.4. Uniqueness:

Let X, Y and Z be Banach spaces and  $\phi \in L^2(X, Y; Z)$ . Suppose that  $(Z, \phi)$  is a tensor product of X with Y. Then  $(Z, \phi)$  is unique up to isomorphism.

Proof. Let W be a Banach space and  $\psi \in L^2(X, Y; W)$ . Suppose that  $(W, \psi)$  is a tensor product of X with Y. Since  $(Z, \phi)$  is a tensor product of X with Y, there exists a unique  $\psi' \in L(Z, W)$  such that  $\psi' \circ \phi = \psi$ . Since  $(W, \psi)$  is a tensor product of X with Y, there exists a unique  $\phi' \in L(W, Z)$  such that  $\phi' \circ \psi = \phi$ . Thus the following diagram commutes:

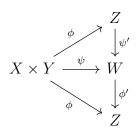


On the other hand, since  $(W, \psi)$  is a tensor product of X with Y, there exists a unique  $\Psi \in L(W)$  such that  $\Psi \circ \psi = \psi$ . Thus the following diagram commutes:

$$\begin{array}{ccc} X\times Y \xrightarrow{\psi} W \\ & \downarrow_{\Psi} \\ & W \end{array}$$

Since  $I_W \in L(W)$  and  $I_W \circ \psi = \psi$ , uniqueness of  $\Psi$  implies that  $\Psi = I_W$ . From the first diagram, we see that  $\psi' \circ \phi'$  satisfies  $(\psi' \circ \phi') \circ \psi = \psi$ . Since  $\psi' \circ \phi' \in L(W)$ , uniqueness of  $\Psi$  implies that  $\Psi = \psi' \circ \phi'$ . Thus  $\psi' \circ \phi' = I_W$ .

Similarly, we could have initially considered the following diagram:



Playing a similar game, we could use the fact that there exists a unique  $\Phi \in L(Z)$  such that  $\Phi \circ \phi = \phi$  to obtain the following diagram:

$$X \times Y \xrightarrow{\phi} Z$$

$$\downarrow^{\Phi}$$

$$Z$$

As before, uniqueness enables us to conclude that  $\phi' \circ \psi' = I_Z$ . Thus  $\psi'$  and  $\phi'$  are isomorphisms and  $Z \cong W$ .

**Note 4.7.5.** The following definitions and exercises will cover the explicit construction of a tensor product of Banach spaces.

**Definition 4.7.6.** Let X and Y be Banach spaces. Define  $X \otimes^{\text{alg}} Y = \text{span}\{x \otimes y : x \in X \text{ and } y \in Y\}$  to be the algebraic tensor product of X with Y (see section ??? of [1] for details).

**Exercise 4.7.7.** Let X and Y be Banach spaces and  $x \otimes y \in X \otimes^{\text{alg}} Y$ . If for each  $\phi \in X^*$  and  $\psi \in Y^*$ ,  $\phi \otimes \psi(x,y) = 0$ , then  $x \otimes y = 0$ .

*Proof.* The previous section tells us that for each  $\phi \in X^*$  and  $\psi \in Y^*$ ,  $\phi \otimes psi(x,y) = 0$ , then x = 0 or y = 0. This implies that  $x \otimes y = 0$ .

## Definition 4.7.8. The Projective Norm:

Define  $\|\cdot\|_{\pi}: X \otimes^{\operatorname{alg}} Y \to [0, \infty)$  by

$$||u||_{\pi} = \inf \left\{ \sum_{j=1}^{n} ||x_{j}|| ||y_{j}|| : (x_{j})_{j=1}^{n} \subset X, (y_{j})_{j=1}^{n} \subset Y \text{ and } u = \sum_{j=1}^{n} x_{j} \otimes y_{j} \right\}$$

**Exercise 4.7.9.** Let X and Y be Banach spaces. Then  $\|\cdot\|_{\pi}: X \otimes^{\operatorname{alg}} Y \to [0, \infty)$  is a norm on  $X \otimes^{\operatorname{alg}} Y$ .

Proof.

• Let  $\lambda \in \mathbb{C}$ ,  $u \in X \otimes^{\operatorname{alg}} Y$ . If  $\lambda = 0$ , then  $\lambda u = 0u = 0 \otimes 0$  and clearly  $\|\lambda u\|_{\pi} = 0 = \|\lambda\| \|u\|_{\pi}$ . Suppose that  $\lambda \neq 0$ . Let  $\epsilon > 0$ . Then there exist  $(x_j)_{j=1}^n \subset X$  and  $(y_j)_{j=1}^n \subset Y$  such that  $u = \sum_{j=1}^n x_j \otimes y_j$  and  $\sum_{j=1}^n \|x_j\| \|y_j\| < \|u\|_{\pi} + \epsilon/|\lambda|$ . Then  $\lambda u = \sum_{j=1}^n (\lambda x_j) \otimes y_j$ . Therefore

$$\|\lambda u\|_{\pi} \leq \sum_{j=1}^{n} \|\lambda x_{j}\| \|y_{j}\|$$

$$\leq |\lambda| \sum_{j=1}^{n} \|x_{j}\| \|y_{j}\|$$

$$< |\lambda| \left( \|u\|_{\pi} + \frac{\epsilon}{|\lambda|} \right)$$

$$= |\lambda| \|u\|_{\pi} + \epsilon$$

Since  $\epsilon > 0$  is arbitrary,  $\|\lambda u\|_{\pi} \leq |\lambda| \|u\|_{\pi}$ . For the sake of contradiction, suppose that  $\|\lambda u\|_{\pi} < |\lambda| \|u\|_{\pi}$ . Then there exists  $(x_j)_{j=1}^n \subset X$  and  $(y_j)_{j=1}^n \subset Y$  such that  $\lambda u = \sum_{j=1}^{n} x_{j} \otimes y_{j} \text{ and } \sum_{j=1}^{n} \|x_{j}\| \|y_{j}\| < |\lambda| \|u\|_{\pi}. \text{ Hence } u = \sum_{j=1}^{n} (\lambda^{-1} x_{j}) \otimes y_{j}. \text{ This implies}$ that

$$||u||_{\pi} \leq \sum_{j=1}^{n} ||\lambda^{-1}x_{j}|| ||y_{j}||$$

$$= |\lambda|^{-1} \sum_{j=1}^{n} ||x_{j}|| ||y_{j}||$$

$$< |\lambda|^{-1} |\lambda| ||u||_{\pi}$$

$$= ||u||_{\pi}$$

which is a contradiction. Therefore  $\|\lambda u\|_{\pi} \geq |\lambda| \|u\|_{\pi}$  which implies that  $\|\lambda u\|_{\pi} =$ 

• Let  $u, v \in X \otimes^{\operatorname{alg}} Y$  and  $\epsilon > 0$ . Then there exist  $(x_j)_{j=1}^n$ ,  $(a_k)_{k=1}^m \subset X$  and  $(y_j)_{j=1}^n$ ,  $(b_k)_{k=1}^m \subset Y$  such that  $u = \sum_{j=1}^n x_j \otimes y_j, \ v = \sum_{k=1}^m a_k \otimes b_k, \ \sum_{j=1}^n \|x_j\| \|y_j\| < \|u\|_{\pi} + \epsilon/2$  and  $\sum_{k=1}^{m} \|a_k\| \|b_k\| < \|u\|_{\pi} + \epsilon/2$ . Then  $u + v = \sum_{j=1}^{n} x_j \otimes y_j + \sum_{k=1}^{m} a_k \otimes b_k$  which implies that

$$||u+v||_{\pi} \le \sum_{j=1}^{n} ||x_{j}|| ||y_{j}|| + \sum_{k=1}^{m} ||a_{k}|| ||b_{k}||$$

$$< ||u||_{\pi} + \epsilon/2 + ||v||_{\pi} + \epsilon/2$$

$$= ||u||_{\pi} + ||v||_{\pi} + \epsilon$$

Since  $\epsilon > 0$  is arbitrary,  $||u+v||_{\pi} \le ||u||_{\pi} + ||v||_{\pi}$ . • Let  $u \in X \otimes^{\text{alg}} Y$ . Suppose that ||u|| = 0. Let  $\phi \in X^*$  and  $\psi \in Y^*$  and  $\epsilon > 0$ . Then there exist  $(x_j)_{j=1}^n \subset X$  and  $(y_j)_{j=1}^n \subset Y$  such that  $u = \sum_{j=1}^n x_j \otimes y_j$  and

$$\sum_{j=1}^{n} \|x_j\| \|y_j\| < \frac{\epsilon}{\|\phi\| \|\psi\| + 1}$$

Then

$$\sum_{j=1}^{n} |\phi \otimes \psi(x_j, y_j)| = \sum_{j=1}^{n} |\phi(x_j)\psi(y_j)|$$

$$\leq \sum_{j=1}^{n} ||\phi|| ||x_j|| ||\psi|| ||y_j||$$

$$= ||\phi|| ||\psi|| \sum_{j=1}^{n} ||x_j|| ||y_j||$$

$$< ||\phi|| ||\psi|| \frac{\epsilon}{||\phi|| ||\psi|| + 1}$$

$$< \epsilon$$

Then for each  $j \in \{1, ..., n\}$ ,  $|\phi \otimes \psi(x_j, y_j)| < \epsilon$ . **FINISH!!!** Try using sequences and continuity and a common refinement of representation and averaging

# Exercise 4.7.10. Existence:

Proof.

# 4.8. The Baire Category and Closed Graph Theorems.

**Theorem 4.8.1.** Let X, Y be Banach spaces and  $T \in L(X, Y)$ . If T is surjective, then T is open.

Corollary 4.8.2. Let X, Y be Banach spaces and  $T \in L(X, Y)$ . If T is a bijection, then  $T^{-1} \in L(X, Y)$ .

**Definition 4.8.3.** Let X, Y be sets and  $f: X \to Y$ . We define the **graph of f**,  $\Gamma(f)$ , by  $\Gamma(f) = \{(x, y) \in X \times Y : f(x) = y\}$ .

**Theorem 4.8.4.** Let X, Y be Banach spaces and  $T: X \to Y$  a linear map. If  $\Gamma(T)$  is closed, then  $T \in L(X, Y)$ .

**Note 4.8.5.** We recall that  $\Gamma(T)$  is closed iff for each  $(x_n)_{n\in\mathbb{N}}\subset X,\ x\in X$  and  $y\in Y,$   $x_n\to x$  and  $T(x_n)\to y$  implies that T(x)=y.

**Theorem 4.8.6.** Let X, Y be Banach spaces and  $S \subset L(X, Y)$ . If for each  $x \in X$ ,

$$\sup_{T \in S} \|Tx\| < \infty$$

then

$$\sup_{T \in S} \|T\| < \infty$$

**Exercise 4.8.7.** Let  $\mu$  be counting measure on  $(N, \mathcal{P}(\mathbb{N}))$ . Define  $h : \mathbb{N} \to \mathbb{N}$  and  $\nu$  on  $(N, \mathcal{P}(\mathbb{N}))$  by h(n) = n and  $d\nu = hd\mu$ . Define  $X = L^1(\nu)$  and  $Y = L^1(\mu)$ . Equip both X and Y with the  $L^1$  norm with respect to  $\mu$ .

- (1) We have that X is a proper subspace of Y and therefore X is not complete.
- (2) Define  $T: X \to Y$  by Tf(n) = nf(n). Then T is linear,  $\Gamma(T)$  is closed, and T is unbounded.
- (3) Define  $S: Y \to X$  by  $Sg(n) = \frac{1}{n}g(n)$ . Then  $S \in L(Y,X)$ , S is surjective and S is not open.

Proof.

(1) Note that for each  $f: \mathbb{N} \to \mathbb{C}$ ,

$$||f||_{\mu,1} = \sum_{n=1}^{\infty} |f(n)|$$

$$\leq \sum_{n=1}^{\infty} n|f(n)|$$

$$= ||f||_{\nu,1}$$

Hence X is a subspace of Y. Define  $f: \mathbb{N} \to \mathbb{C}$  by  $f(n) = \frac{1}{n^2}$ . Then

$$||f||_{\mu,1} = \sum_{n=1}^{\infty} \frac{1}{n^2} < \infty$$

So  $f \in Y$ . However

$$||f||_{\nu,1} = \sum_{n=1}^{\infty} \frac{1}{n} = \infty$$

So  $f \notin X$ . Thus X is a proper subspace of Y. Let  $g \in Y$  and  $\epsilon > 0$ . Since the simple functions are dense in  $L^1(\mu)$ , there exists  $\phi \in L^1(\mu)$  such that  $\phi$  is simple and  $\|g - \phi\|_{\mu,1} < \epsilon$ . Then there exist  $(c_i)_{i=1}^k \subset \mathbb{C}$  and  $(E_i)_{i=1}^k \subset \mathcal{P}(\mathbb{N})$  such that for each  $i, j \in \{1, 2, \dots, k\}$ ,  $E_i$  is finite,  $i \neq j$  implies that  $E_i \cap E_j = \emptyset$  and

$$\phi = \sum_{i=1}^{k} c_i \chi_{E_i}$$

Define  $c = \max\{|c_i| : i = 1, 2, \dots k\}$  and  $m = \max\left[\bigcup_{i=1}^k E_i\right]$ . Then

$$\|\phi\|_{\nu,1} = \sum_{n=1}^{m} n|\phi(n)|$$

$$\leq \sum_{n=1}^{m} mc$$

$$= cm^{2}$$

$$\leq \infty$$

Hence  $\phi \in X$  and X is dense in Y. Since X is a dense, proper subspace, it is not closed. Since Y is complete and  $X \subset Y$  is not closed, we have that X is not complete.

(2) Clearly T is linear. Let  $(f_j)_{j\in\mathbb{N}}\subset X$ ,  $f\in X$  and  $g\in Y$ . Suppose that  $f_j\xrightarrow{L^1(\mu)} f$  and  $Tf_j\xrightarrow{L^1(\mu)} g$ .

Note that for each  $j \in \mathbb{N}$  and  $n \in \mathbb{N}$ ,

$$|f_j(n) - f(n)| \le \sum_{n=1}^{\infty} |f_j(n) - f(n)| = ||f_j - f||_{\mu,1}$$

and

$$|nf_j(n) - g(n)| \le \sum_{n=1}^{\infty} |nf_j(n) - g(n)| = ||Tf_j - g||_{\mu,1}$$

Thus for each  $n \in \mathbb{N}$ ,  $f_j(n) \xrightarrow{j} f(n)$  and  $nf_j(n) \xrightarrow{j} g(n)$ . This implies that for each  $n \in \mathbb{N}$ , nf(n) = g(n). Thus Tf = g which implies that  $\Gamma(T)$  is closed. Suppose, for the sake of contradiction, that T is bounded. Then there exists  $C \geq 0$  such that for each  $f \in X$ ,  $||Tf||_{\mu,1} \leq C||f||_{\mu,1}$ . Choose  $n \in \mathbb{N}$  such that n > C. Define  $f : \mathbb{N} \to \mathbb{C}$  by  $f = \chi_{\{n\}}$ . As established above,  $S^+ \subset L^1(\mu)$ . Then  $||f||_{\mu,1} = 1$  and

$$||Tf||_{\mu,1} = n$$
  
>  $C$   
=  $C||f||_{\mu,1}$ 

which is a contradiction. So T is unbounded.

(3) Clearly S is linear. Let  $q \in Y$ . Then

$$||Sg||_{\mu,1} = \sum_{n=1}^{\infty} \frac{1}{n} |g(n)|$$

$$\leq \sum_{n=1}^{\infty} |g(n)|$$

$$= ||g||_{\mu,1}$$

So S is bounded and  $||S|| \le 1$ . Thus  $S \in L(Y, X)$ . Let  $f \in X$ . Define  $g : \mathbb{N} \to \mathbb{C}$  by g(n) = nf(n). By definition,  $g \in Y$  and we have that

$$Sg(n) = \frac{1}{n}g(n)$$
$$= f(n)$$

Hence Sg = f and thus S is surjective. Let  $g \in Y$ . Suppose that Sg = 0. Then

$$\sum_{n=1}^{\infty} \frac{1}{n} |g(n)| = ||Sg|| = 0$$

Thus for each  $n \in \mathbb{N}$ , g(n) = 0. Hence  $\ker S = \{0\}$  and S is injective. Note that for each  $A \subset Y$ ,  $S(A) = T^{-1}(A)$ . If S is open, then T is continuous which as shown above is a contradiction. So g is not open.

**Exercise 4.8.8.** Let  $X = C^1([0,1])$  and Y = C([0,1]). Equip both X and Y with the uniform norm.

- (1) Then X is not complete
- (2) Define  $T: X \to Y$  by Tf = f'. Then  $\Gamma(T)$  is closed and T is not bounded.

*Proof.* (1) Recall that for each  $a, b \ge 0$  and  $p \in \mathbb{N}$ ,

$$(a^{\frac{1}{p}} + b^{\frac{1}{p}})^p = \sum_{n=0}^p \binom{p}{n} a^{\frac{n}{p}} b^{\frac{p-n}{p}} \ge a+b$$

Thus  $(a+b)^{\frac{1}{p}} \le a^{\frac{1}{p}} + b^{\frac{1}{p}}$ .

For each  $n \in \mathbb{N}$ , define  $f_n : [0,1] \to \mathbb{C}$  by  $f_n(x) = \sqrt{(x-\frac{1}{2})^2 + \frac{1}{n^2}}$ . Then  $(f_n)_{n \in \mathbb{N}} \subset X$ . Define  $f : [0,1] \to \mathbb{C}$  by  $f(x) = |x-\frac{1}{2}|$ . Then  $f \in Y \cap X^c$ . Note that for each  $n \in \mathbb{N}$ ,  $f \leq f_n$ . Our observation above implies that for each  $x \in X$ ,

$$f_n(x) = \left[ (x - \frac{1}{2})^2 + \frac{1}{n^2} \right]^{\frac{1}{2}}$$

$$\leq |x - \frac{1}{2}| + \frac{1}{n}$$

Thus  $0 \le f_n - f \le \frac{1}{n}$ . This implies that  $f_n \xrightarrow{\mathrm{u}} f$ . Since  $f \notin X$ , X is not complete.

(2) Let  $(f_n)_{n\in\mathbb{N}}\subset X$ ,  $f\in X$  and  $g\in Y$ . Suppose that  $f_n\stackrel{\mathrm{u}}{\to} f$  and  $Tf_n\stackrel{\mathrm{u}}{\to} g$ . Let  $x\in[0,1]$ . Then  $f_n(x)\to f(x)$  and  $f_n(0)\to f(0)$  and  $f_n'\stackrel{\mathrm{u}}{\to} g$ . Applying the DCT to this sequence of integrable functions that converges uniformly to an integrable function on a finite measure space (a previous exercise) we have that

$$f_n(x) - f_n(0) = \int_{[0,x]} f'_n dm$$
$$\to \int_{[0,x]} g dm$$

Since  $f_n(x) - f_n(0) \to f(x) - f(0)$ , we know that

$$f(x) - f(0) = \int_{[0,x]} gdm$$

Thus Tf = g and  $\Gamma(T)$  is closed.

By Exercise 4.2.2, T is not bounded.

**Exercise 4.8.9.** Let X, Y be Banach spaces and  $T \in L(X, Y)$ . Then  $X/\ker T \cong T(X)$  iff T(X) is closed.

*Proof.* Since X is a banach space and T is continuous, we have that  $\ker T$  is closed and  $X/\ker T$  is a Banach space. Suppose that  $X/\ker T\cong T(X)$ . Then T(X) is complete. Since Y is complete, this implies that T(X) is closed.

Conversely Suppose that T(X) is closed. Then T(X) is complete. Define  $S: X/\ker T \to T(X)$  by  $S(x+\ker T)=T(x)$ . A previous exercise tells us that the map  $S: X/\ker T \to T(X)$  defined by  $S(x+\ker T)=T(x)$  is a bounded linear bijection. Since T(X) is complete and S is surjective,  $S^{-1}$  is bounded and thus S is an isomorphism.

**Exercise 4.8.10.** Let X be a separable Banach space. Define  $B_X = \{x \in X : ||x|| < 1\}$ . Let  $(x_n)_{n \in \mathbb{N}} \subset B_X$  a dense subset of the unit ball and  $\mu$  the counting measure on  $(\mathbb{N}, \mathcal{P}(\mathbb{N}))$ . Define  $T : L^1(\mu) \to X$  by

$$Tf = \sum_{n=1}^{\infty} f(n)x_n$$

Then

- (1) T is well defined and  $T \in L(L^1(\mu), X)$
- (2) T is surjective
- (3) There exists a closed subspace  $K \subset L^1(\mu)$  such that  $L^1(\mu)/K \cong X$

*Proof.* (1) Let  $f \in L^1(\mu)$ . Since X is complete and

$$\sum_{n=1}^{\infty} ||f(n)x_n|| = \sum_{n=1}^{\infty} |f(n)|||x_n||$$

$$\leq \sum_{n=1}^{\infty} |f(n)|$$

$$< \infty$$

we have that  $\sum_{n=1}^{\infty} f(n)x_n$  converges and thus  $Tf \in X$ . Hence T is well defined.

Clearly T is linear. Let  $f \in L^1(\mu)$ . Then

$$||Tf|| = ||\sum_{n=1}^{\infty} f(n)x_n||$$

$$\leq \sum_{n=1}^{\infty} ||f(n)x_n||$$

$$\leq \sum_{n=1}^{\infty} |f(n)||$$

$$= ||f||_1$$

So T is bounded with  $||T|| \leq 1$ .

(2) Let  $x \in X$ . Suppose that ||x|| < 1. Then  $x \in B_X$ . So there exists  $n_1 \in \mathbb{N}$  such that  $||x - x_{n_1}|| < \frac{1}{2}$ . Then  $2(x - x_{n_1}) \in B_X$ . Since for each  $j \in \mathbb{N}$ ,  $B_X \setminus (x_n)_{n=1}^j$  is dense in  $B_X$ , there exists  $n_2 \in \mathbb{N}$  such that  $x_{n_2} \notin (x_n)_{n=1}^{n_1}$  and  $||2(x - x_{n_1}) - x_{n_2}|| < \frac{1}{2}$  which implies that  $||x - (x_{n_1} - \frac{1}{2}x_{n_2})|| < \frac{1}{4}$ .

Proceed inductively to obtain a subsequence  $(x_{n_k})_{k\in\mathbb{N}}$  such that for each  $k\geq 2$ ,  $x_{n_k} \notin (x_n)_{n=1}^{n_{k-1}}$  and  $||x-\sum_{j=1}^k 2^{1-j}x_{n_j}|| < \frac{1}{2^k}$ . Then  $x=\sum_{k=1}^\infty 2^{1-k}x_{n_k}$ .

Define  $f: \mathbb{N} \to \mathbb{C}$  by  $f = \sum_{k=1}^{\infty} 2^{1-k} \chi_{\{n_k\}}$ . Then  $||f||_1 = \sum_{k=1}^{\infty} 2^{1-k} < \infty$ , so  $f \in L^1(\mu)$  and  $Tf = \sum_{k=1}^{\infty} 2^{1-k} x_{n_k} = x$ . Now, suppose that  $||x|| \ge 1$ , then  $\frac{1}{2||x||} x \in B_X$ . The above argument shows that there exists  $f \in L^1(\mu)$  such that  $Tf = \frac{1}{2||x||} x$ . Then  $2||x||f \in L^1(\mu)$  and T(2||x||f) = 2||x||Tf = x.

So for each  $x \in X$ , there exists  $f \in L^1(\mu)$  such that Tf = x and thus T is surjective.

(3) Since X is a Banach space and T is surjective, the previous exercise implies that  $L^1(\mu)/\ker T \cong X$ .

**Exercise 4.8.11.** Let X, Y be Banach spaces and  $T: X \to Y$  a linear map. If for each  $f \in Y^*$ ,  $f \circ T \in X^*$ , then  $T \in L(X, Y)$ .

*Proof.* Suppose that for each  $f \in Y^*$ ,  $f \circ T \in X^*$ . Let  $x \in X$ ,

# 4.9. Banach Algebras.

**Definition 4.9.1.** Let X be a Banach space and an associative algebra. Then X is said to be a **Banach algebra** if for each  $S, T \in X$ ,  $||ST|| \le ||S|| ||T||$ .

**Definition 4.9.2.** Let X be a Banach algebra and  $I \in X$ . Then I is said to be an **identity** if for each  $T \in X$ , IT = TI = T.

**Definition 4.9.3.** Let X be a Banach algebra. and  $I \in X$ . Then I is said to be an **identity** if  $I \neq 0$  and for each  $T \in X$ , IT = TI = T.

**Definition 4.9.4.** Let X be a Banach algebra. Then X is said to be **unital** if there exists  $I \in X$  such that I is an identity.

**Exercise 4.9.5.** Let X be a unital Banach algebra. Then there exists a unique  $I \in X$  such that I is an identity.

Proof. Clear.  $\Box$ 

Note 4.9.6. We denote the unique identity element by I.

**Definition 4.9.7.** Let X be a unital Banach algebra and  $T, S \in X$ . Then S is said to be an **inverse** of T if TS = ST = I.

**Definition 4.9.8.** Let X be a unital Banach algebra and  $T \in X$ . Then T is said to be **invertible** if there exists  $S \in X$  such that S is an inverse of T.

**Exercise 4.9.9.** Let X be a unital Banach algebra and  $T \in X$ . If T is invertible, then there exists a unique  $S \in X$  such that S is an inverse of T.

Proof. Clear.  $\Box$ 

Note 4.9.10. We denote the unique inverse of T by  $T^{-1}$ .

#### Exercise 4.9.11. Fundamental Example:

Let X be a Banach space. Then GL(X) is a unital Banach algebra.

Proof. Clear.  $\Box$ 

**Definition 4.9.12.** Let X be a unital Banach algebra. We define  $GL(X) = \{T \in X : T \text{ is invertible}\}.$ 

**Exercise 4.9.13.** Let X be a unital Banach algebra. Then GL(X) is a group.

Proof. Clear.  $\Box$ 

**Exercise 4.9.14.** Let X be a unital Banach algebra. Then  $1 \leq ||I||$ .

*Proof.* Since  $I \neq 0$ ,  $||I|| \neq 0$ . By definition,

$$||I|| = ||II|| \le ||I|| ||I||$$

Hence  $1 \leq ||I||$ .

**Exercise 4.9.15.** Let X be a Banach algebra. Then mulitplication is continuous.

*Proof.* Let  $(S_1, T_1) \in X \times X$  and  $\epsilon > 0$ . Choose  $\delta = \min\{\frac{\epsilon}{2(\|S_1\| + \|T_1\| + 1)}, \frac{\sqrt{\epsilon}}{\sqrt{2}}\}$ . Let  $(S_2, T_2) \in X \times X$ . Suppose that

$$||(S_1, T_1) - (S_2, T_2)|| = \max\{||S_2 - S_2||, ||T_1 - T_2||\} < \delta$$

Then

$$||S_1T_1 - S_2T_2|| = ||S_1T_1 - S_2T_1 + S_2T_1 - S_2T_2||$$

$$\leq ||S_1 - S_2|| ||T_1|| + ||S_2|| ||T_1 - T_2||$$

$$\leq ||S_1 - S_2|| ||T_1|| + (||S_1 - S_2|| + ||S_1||) ||T_1 - T_2||$$

$$\leq \delta ||T_1|| + (\delta + ||S_1||) \delta$$

$$= \delta (||S_1|| + ||T_1||) + \delta^2$$

$$< \frac{\epsilon}{2} + \frac{\epsilon}{2}$$

$$= \epsilon$$

**Exercise 4.9.16.** Let X be a unital Banach algebra. Then

(1) For each  $T \in X$ , if ||I - T|| < 1, then  $T \in GL(X)$  and

$$T^{-1} = \sum_{n=0}^{\infty} (I - T)^n$$

- (2) For each  $S, T \in X$ , if  $S \in GL(X)$  and  $||S T|| < ||S^{-1}||^{-1}$ , then  $T \in GL(X)$ .
- (3) GL(X) is open.

Proof.

(1) Let  $T \in X$ . Suppose that ||I - T|| < 1. Then

$$\sum_{n=0}^{\infty} \|(I-T)^n\| \le \sum_{n=0}^{\infty} \|I-T\|^n < \infty$$

Since X is a complete,  $\sum_{n=0}^{\infty} (I-T)^n$  converges in X.

Define 
$$(S_k)_{k=0}^{\infty} \subset X$$
 and  $S \in X$  by  $S_k = \sum_{n=0}^{k} (I-T)^n$  and

$$S = \sum_{n=0}^{\infty} (I - T)^n$$
. Then for each  $k \in \mathbb{N}$ ,

$$S_k T = S_k - S_k (I - T)$$
  
=  $(I - T)^0 - (I - T)^{k+1}$   
=  $I - (I - T)^{k+1}$ 

and  $||S_kT - I|| \le ||I - T||^{k+1}$ . Since multiplication on Banach algebras is continuous, we have that

$$ST = (\lim_{k \to \infty} S_k)T = \lim_{k \to \infty} S_kT = I$$

Similarly TS = I. Thus  $T \in GL(X)$  and  $T^{-1} = S \in X$ .

(2) Let 
$$S,T\in X$$
. Suppose that  $S\in GL(X)$  and  $\|S-T\|<\|S^{-1}\|^{-1}$ . Then 
$$\|I-S^{-1}T\|=\|S^{-1}(S-T)\|$$
 
$$\leq \|S^{-1}\|\|S-T\|$$
 
$$<1$$
 So  $S^{-1}T\in GL(X)$ . Thus  $T=S(S^{-1}T)\in GL(X)$ .

(3) Let 
$$T \in GL(X)$$
. Choose  $\delta = \|T^{-1}\|^{-1}$ . By (2),  $B(T, \delta) \subset GL(X)$ .

#### 5. Hilbert Spaces

### 5.1. Introduction.

**Definition 5.1.1.** Let H be a vector space and  $\langle \cdot, \cdot \rangle : H \to \mathbb{C}$ . Then  $\langle \cdot, \cdot \rangle$  is said to be an inner product on H if for each  $x, y, z \in H$  and  $c \in \mathbb{C}$ 

- (1)  $\langle x, y + cz \rangle = \langle x, y \rangle + c \langle x, z \rangle$
- $(2) \langle x, y \rangle = \langle y, x \rangle^*$
- (3)  $\langle x, x \rangle \geq 0$
- (4) if  $\langle x, x \rangle = 0$ , then x = 0.

**Note 5.1.2.** In mathematics, inner products are conventionally defined to be linear in the first argument. However, in my opinion, the convention in physics of defining inner products to be linear in the second argument makes more sense.

**Exercise 5.1.3.** Let H be an inner product space,  $(x_j)_{j=1}^n$ ,  $(y_j)_{j=1}^n \subset H$  and  $(\alpha_j)_{j=1}^n$ ,  $(\beta_j)_{j=1}^n \subset \mathbb{C}$ . Then

$$\left\langle \sum_{i=1}^{n} \alpha_i x_i, \sum_{j=1}^{n} \beta_j y_j \right\rangle = \sum_{i=1}^{n} \sum_{j=1}^{n} \alpha_i^* \beta_j \langle x_i, y_j \rangle$$

Proof. Clear.

**Definition 5.1.4.** Let H be an inner product space. Define the **induced norm**, denoted  $\|\cdot\|: H \to \mathbb{C}$ , by

$$||x|| = \langle x, x \rangle^{\frac{1}{2}}$$

#### Exercise 5.1.5. Cauchy-Schwarz Inequality

Let H be an inner product space. Then for each  $x, y \in H$ ,  $|\langle x, y \rangle| \leq ||x|| ||y||$  and  $|\langle x, y \rangle| = ||x|| ||y||$  iff  $x \in \text{span}(y)$ .

**Hint:** For  $x, y \in H$ , put  $z = \operatorname{sgn}\langle x, y \rangle^* y$  and Consider  $f : \mathbb{R} \to [0, \infty)$  defined by  $f(t) = \|x - tz\|^2$ 

*Proof.* Let  $x, y \in H$ . If y = 0, then the claim holds trivially. Suppose that  $y \neq 0$ . Put  $z = \operatorname{sgn}\langle x, y \rangle^* y$ . So  $\langle x, z \rangle = |\langle x, y \rangle|$  and ||z|| = ||y||. Define  $f : \mathbb{R} \to [0, \infty)$  by

$$f(t) = ||x - tz||^2$$

. Then for each  $t \in \mathbb{R}$ ,

$$0 \le f(t)$$

$$= ||x - tz||^{2}$$

$$= ||x||^{2} + |t|^{2}||z||^{2} - 2\operatorname{Re}(t\langle x, z\rangle)$$

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

Thus f is a quadratic with a minimum at  $t_0 = \frac{|\langle x, y \rangle|}{||y||^2}$ . Hence

$$0 \le f(t_0)$$

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

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

Which implies that

$$|\langle x, y \rangle|^2 \le ||x||^2 ||y||^2$$

and hence the claim holds. Clearly if  $x \in \text{span}(y)$ , then equality holds. Conversely, if equality holds, then x - z = 0 which implies that  $x \in \text{spn}(y)$ .

**Exercise 5.1.6.** Let H be an inner product space. Then the induced norm,  $\|\cdot\|: H \to \mathbb{C}$ , is a norm.

*Proof.* Let  $x, y \in H$  and  $c \in \mathbb{C}$ . Then

- (1) By definition, if ||x|| = 0, then  $\langle x, x \rangle = 0$ , which implies that x = 0.
- (2) Note that

$$||cx||^2 = \langle cx, cx \rangle$$
$$= c * c \langle x, x \rangle$$
$$= |c|^2 ||x||^2$$

So ||cx|| = |c|||x||

(3) The Cauchy-Schwarz inequality implies that

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

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

$$\leq ||x||^2 + ||y||^2 + 2||x|| ||y||$$

$$= (||x|| + ||y||)^2$$

Hence  $||x + y|| \le ||x|| + ||y||$ .

**Definition 5.1.7.** Let H be an inner product space,  $x, y \in H$  and  $S \subset H$ . Then

- (1) x and y are said to be **orthogonal** if  $\langle x, y \rangle = 0$ .
- (2) S is said to be **orthogonal** if for each  $x, y \in S$ , x, y are orthogonal.

# Exercise 5.1.8. (Pythagorean theorem):

Let H be an inner product space and  $(x_j)_{j=1}^n \subset H$  an orthogonal set. Then

$$\left\| \sum_{j=1}^{n} x_j \right\|^2 = \sum_{j=1}^{n} \|x_j\|^2$$

*Proof.* We have that

$$\left\| \sum_{j=1}^{n} x_j \right\|^2 = \left\langle \sum_{i=1}^{n} x_i, \sum_{j=1}^{n} x_j \right\rangle$$

$$= \sum_{i=1}^{n} \sum_{j=1}^{n} \langle x_j, x_j \rangle$$

$$= \sum_{j=1}^{n} \langle x_j, x_j \rangle$$

$$= \sum_{j=1}^{n} \|x_j\|^2$$

**Exercise 5.1.9.** Let H be an inner product space and  $S \subset H$ . Suppose that  $0 \notin S$ . If S is orthogonal, then S is linearly independent.

*Proof.* Let  $x_1, \dots, x_n \in S$  and  $c_1, \dots, c_n \in \mathbb{C}$ . Suppose that  $\sum_{j=1}^n c_j x_j = 0$ . Since  $(c_j x_j)_{j=1}^n$  is orthogonal, the Pythagorean theorem implies that

$$0 = \left\| \sum_{i=1}^{n} c_i x_i \right\|$$
$$= \sum_{j=1}^{n} |c_j|^2 \|x_j\|$$

So for each  $j \in \{1, \dots, n\}$ ,  $c_j = 0$  and S is linearly independent.

**Definition 5.1.10.** Let H be an inner product space and  $S \subset H$ . Then S is said to be **orthonormal** if S is orthogonal and for each  $x \in S$ , ||x|| = 1.

#### Exercise 5.1.11. Bessel's Inequality:

Let H be an inner product space and  $S \subset H$ . If S is orthonormal, then for each  $x \in H$ ,

$$\sum_{u \in S} |\langle u, x \rangle|^2 \le ||x||$$

and in particular,  $\{u \in S : \langle u, x \rangle \neq 0\}$  is countable.

*Proof.* Suppose that S is orthonormal. Let  $x \in H$  and  $F \subset S$  finite. Then the Pythagorean theorem implies that

$$0 \le \left\| x - \sum_{u \in F} \langle u, x \rangle u \right\|^2$$

$$= \|x\|^2 + \left\| \sum_{u \in F} \langle u, x \rangle u \right\|^2 - 2 \operatorname{Re} \left\langle x, \sum_{u \in F} \langle u, x \rangle u \right\rangle$$

$$= \|x\|^2 + \sum_{u \in F} |\langle u, x \rangle|^2 \|u\|^2 - 2 \sum_{u \in F} |\langle u, x \rangle|^2$$

$$= \|x\|^2 - \sum_{u \in F} |\langle u, x \rangle|^2$$

So

$$\sum_{u \in F} |\langle u, x \rangle|^2 \le ||x||$$

By definition of the sum,

$$\sum_{u \in S} |\langle u, x \rangle|^2 \le ||x||$$

Basic integration theory then tells us that  $\{u \in S : \langle u, x \rangle \neq 0\}$  is countable.

**Definition 5.1.12.** Let H be an inner product space. Then H is said to be a **Hilbert space** if H is a complete with respect to the induced norm on H.

**Exercise 5.1.13.** Let H be a Hilbert space and  $S \subset H$ . Suppose that S is orthonormal. Then the following are equivalent:

- (1) For each  $x \in H$ , if for each  $u \in S$ ,  $\langle u, x \rangle = 0$ , then x = 0.
- (2) For each  $x \in H$ , there exist  $(u_j)_{j \in \mathbb{N}} \subset S$  such that  $x = \sum_{j \in \mathbb{N}} \langle u_j, x \rangle u_j$  and for each
- $u \notin (u_j)_{j \in \mathbb{N}}, \langle u, x \rangle = 0.$ (3) For each  $x \in H$ ,  $||x||^2 = \sum_{u \in S} |\langle u, x \rangle|^2$ .

Proof.

 $\bullet$  (1)  $\Longrightarrow$  (2):

Suppose that for each  $x \in H$ , if for each  $u \in S$ ,  $\langle u, x \rangle = 0$ , then x = 0. Let  $x \in H$ . Put  $S_* = \{u \in S : \langle u, x \rangle \neq 0\}$ . The previous exercise implies that  $S_*$  is countable. Write  $S_* = (u_j)_{j=1}^n$ . The previous exercise tells us that  $\sum_{j \in \mathbb{N}} |\langle u_j, x \rangle|^2 \leq ||x||^2$  and hence

converges. Thus for  $\epsilon > 0$ , there exist  $N \in \mathbb{N}$  such that for each  $m, n \in \mathbb{N}, m, n \geq N$  implies that if m < n, then

$$\sum_{m+1}^{n} |\langle u_j, x \rangle|^2 < \epsilon$$

Define  $(y_n)_{n\in\mathbb{N}}\subset H$  by

$$y_n = \sum_{j=1}^n \langle u_j, x \rangle u_j$$

Then for each  $m, n \in \mathbb{N}$ ,  $m, n \geq N$  implies that if m < n, then

$$||y_n - y_m||^2 = \left\| \sum_{1}^{n} \langle u_j, x \rangle u_j - \sum_{1}^{m} \langle u_j, x \rangle u_j \right\|^2$$

$$= \left\| \sum_{m+1}^{n} \langle u_j, x \rangle u_j \right\|^2$$

$$= \sum_{m+1}^{n} |\langle u_j, x \rangle|^2$$

$$< \epsilon$$

So  $(y_n)_{n\in\mathbb{N}}$  is Cauchy. Since H is complete, there exists  $y\in H$  such that  $y_n\to y$ . By definition,

$$y = \sum_{j \in \mathbb{N}} \langle u_j, x \rangle u_j$$

Continuity of  $\langle \cdot, \cdot \rangle : H \times H \to \mathbb{C}$  implies that

(1) for each  $u \in S \setminus S_*$ ,

$$\begin{split} \langle u, x - y \rangle &= \langle u, x \rangle - \langle u, y \rangle \\ &= \langle u, x \rangle - \lim_{n \to \infty} \langle u, y_n \rangle \\ &= \langle u, x \rangle - \lim_{n \to \infty} \sum_{j=1}^{n} \langle u_j, x \rangle \langle u, u_j \rangle \\ &= 0 - 0 \\ &= 0 \end{split}$$

(2) for each  $k \in \mathbb{N}$ ,

$$\langle u_k, x - y \rangle = \langle u_k, x \rangle - \langle u_k, y \rangle$$

$$= \langle u_k, x \rangle - \lim_{n \to \infty} \langle u_k, y_n \rangle$$

$$= \langle u_k, x \rangle - \lim_{n \to \infty} \sum_{j=1}^n \langle u_j, x \rangle \langle u_k, u_j \rangle$$

$$= \langle u_k, x \rangle - \langle u_k, x \rangle$$

$$= 0$$

So for each  $u \in S$ ,  $\langle u, x - y \rangle = 0$ . By assumption, x - y = 0 and hence

$$x = \sum_{j \in \mathbb{N}} \langle u_j, x \rangle u_j$$

• (2)  $\Longrightarrow$  (3): Suppose that for each  $x \in H$ , there exist  $(u_j)_{j \in \mathbb{N}} \subset S$  such that  $x = \sum_{j \in \mathbb{N}} \langle u_j, x \rangle u_j$  and for each  $u \notin (u_j)_{j \in \mathbb{N}}$ ,  $\langle u, x \rangle = 0$ . Then continuity of  $\| \cdot \| : H \to [0, \infty)$  implies that

$$||x||^2 = \left\| \lim_{n \to \infty} \sum_{j=1}^n \langle u_j, x \rangle u_j \right\|^2$$

$$= \lim_{n \to \infty} \left\| \sum_{j=1}^n \langle u_j, x \rangle u_j \right\|^2$$

$$= \lim_{n \to \infty} \sum_{j=1}^n |\langle u_j, x \rangle|^2$$

$$= \sum_{j \in \mathbb{N}} |\langle u_j, x \rangle|^2$$

$$= \sum_{u \in S} |\langle u, x \rangle|^2$$

• (3)  $\Longrightarrow$  (4): Suppose that for each  $x \in H$ ,  $||x||^2 = \sum_{u \in S} |\langle u, x \rangle|^2$ . Let  $x \in H$ . Suppose that for each  $u \in S$ ,  $\langle u, x \rangle = 0$ . Then

$$||x||^2 = \sum_{u \in S} |\langle u, x \rangle|^2$$
$$= 0$$

So x = 0

**Definition 5.1.14.** Let H be a Hilbert space and  $S \subset H$ . Then S is said to be an **orthonormal basis of** H if

- (1) S is orthonormal
- (2) for each  $x \in H$ , if for each  $u \in S$ ,  $\langle u, x \rangle = 0$ , then x = 0

## 5.2. Operators and Functionals.

# Definition 5.2.1. (Adjoint of an Operator):

Let H be a Hilbert space and  $A, B \in L(H)$ . Then B is said to be the **adjoint** of A if for each  $x_1, x_2 \in H$ ,

$$\langle x_1, Ax_2 \rangle = \langle Bx_1, x_2 \rangle$$

In this case, we write

$$B = A^*$$

**Note 5.2.2.** In physics, the adjoint of A is typically denoted by  $A^{\dagger}$ .

**Exercise 5.2.3.** Let H be a Hilbert space,  $A, B \in L(H)$  and  $\lambda \in \mathbb{C}$ , then

- $(1) (A^*)^* = A$
- (2)  $(A+B)^* = A^* + B^*$
- $(3) (AB)^* = B^*A^*$
- $(4) (\lambda A)^* = \lambda^* A^*$
- (5) A and B commute iff  $A^*$  and  $B^*$  commute.

*Proof.* Let  $x_1, x_2 \in H$ . Then

(1)

$$\langle Ax_1, x_2 \rangle = \langle x_2, Ax_1 \rangle^*$$
  
=  $\langle A^*x_2, x_1 \rangle^*$  (by definition)  
=  $\langle x_1, A^*x_2 \rangle$ 

(2)

$$\langle x_1, (A+B)x_2 \rangle = \langle x_1, Ax_2 \rangle + \langle x_1, Bx_2 \rangle$$
$$= \langle A^*x_1, x_2 \rangle + \langle B^*x_1, x_2 \rangle$$
$$= \langle (A^* + B^*)x_1, x_2 \rangle$$

(3)

$$\langle x_1, ABx_2 \rangle = \langle A^*x_1, Bx_2 \rangle$$
  
=  $\langle B^*A^*x_1, x_2 \rangle$ 

(4)

$$\langle x_1, \lambda A x_2 \rangle = \lambda \langle x_1, A x_2 \rangle$$
$$= \lambda \langle A^* x_1, x_2 \rangle$$
$$= \langle \lambda^* A^* x_1, x_2 \rangle$$

(5) If A and B commute, then

$$A^*B^* = (BA)^*$$
$$= (AB)^*$$
$$= B^*A^*$$

Conversely, if  $A^*$  and  $B^*$  commute then

$$AB = (B^*A^*)^*$$
$$= (A^*B^*)^*$$
$$= BA$$

**Definition 5.2.4.** Let H be a Hilbert space and  $Q \in L(H)$ . Then Q is said to be **self-adjoint** if

$$Q = Q^*$$

**Exercise 5.2.5.** Let H be a Hilbert space and  $Q \in L(H)$ . If Q is a self-adjoint then

- (1) the eigenvalues of Q are real.
- (2) the eigenvectors of Q corresponding to distinct eigenvalues are orthogonal.

*Proof.* Suppose that Q is self-adjoint.

(1) Let  $\lambda$  be an eigenvalue of Q with corresponding eigenvector x. Then

$$\lambda \langle x, x \rangle = \langle x, Qx \rangle$$
$$= \langle Qx, x \rangle$$
$$= \lambda^* \langle x, x \rangle$$

Thus  $\lambda = \lambda^*$  and is real

(2) Let  $\lambda_1$  and  $\lambda_2$  be eigenvalues of Q with corresponding eigenvectors  $x_1$  and  $x_2$ . Suppose that  $\lambda_1 \neq \lambda_2$ . Then

$$\lambda_2 \langle x_1, x_2 \rangle = \langle x_1, Qx_2 \rangle$$
$$= \langle Qx_1, x_2 \rangle$$
$$= \lambda_1 \langle x_1, x_2 \rangle$$

So  $(\lambda_2 - \lambda_1)\langle x_1, x_2 \rangle = 0$ . Which implies that  $\langle x_1, x_2 \rangle = 0$ 

**Exercise 5.2.6.** Let H be a Hilbert space,  $A, B \in L(H)$  and  $\lambda \in \mathbb{R}$ . Suppose that A, B are self-adjoint. If A and B commute and then  $\lambda AB$  is self-adjoint.

Proof.

$$(\lambda AB)^* = \lambda^* (AB)^*$$
$$= \lambda B^* A^*$$
$$= \lambda BA$$
$$= \lambda AB$$

Definition 5.2.7. (Adjoint of a Vector):

Let H be a Hilbert space and  $x \in H$ . We define the **adjoint** of x, denoted  $x^* \in H^*$ , by  $x^*y = \langle x, y \rangle$ .

Note 5.2.8. In mathematics, where linearity of the inner product is in the first argument,  $x^*$  is typically referred to by  $u_x \in H^*$  where  $u_x(y) = \langle y, x \rangle$ . In physics, where the inner product with linearity in the second argument,  $x^*\phi$  is usually written in the so-called "bra-ket" notation as  $\langle x|\phi\rangle$  which works smoothly since it aligns with the linearity of  $u_x(\phi_1 + \lambda\phi_2)$  and the conjugate-linearity of  $u_{x_1+\lambda x_2}(\phi)$ . In this way, it generalizes the notation for  $\langle x,y\rangle = x^Ty$  for  $\mathbb{R}^n$  to  $\langle x,y\rangle = x^*y^*$  for  $\mathbb{C}^n$ .

**Exercise 5.2.9.** Let H be a Hilbert space,  $x, y \in H$  and  $\lambda \in \mathbb{C}$ . Then

- $(1) (x+y)^* = x^* + y^*$
- (2)  $(\lambda x)^* = \lambda^* x^*$

Proof. Clear.

**Definition 5.2.10.** Let H be a Hilbert space,  $x, y \in H$  and  $A \in L(H)$ . We define

- (1)  $x^*A \in H^*$  by  $(x^*A)y = x^*(Ay)$
- (2)  $xy^* \in L(H)$  by  $(xy^*)z = (y^*z)x$

**Exercise 5.2.11.** Let H be a Hilbert space,  $A \in L(H)$  and  $x \in H$ . Then

$$(Ax)^* = x^*A^*$$

*Proof.* Let  $y \in H$ . Then

$$(Ax)^*y = \langle Ax, y \rangle$$
$$= \langle x, A^*y \rangle$$
$$= x^*A^*y$$

## Definition 5.2.12. (Commutator):

Let H be a Hilbert space and  $A, B \in L(H)$ . The **commutator** of A and B, denoted [A, B], is defined by

$$[A, B] = AB - BA$$

**Exercise 5.2.13.** Let H be a Hilbert space and  $A, B, C \in L(H)$ . Then

- (1) [AB, C] = A[B, C] + [A, C]B
- (2) [A, BC] = B[A, C] + [A, B]C

Proof.

(1)

$$[AB, C] = ABC - CAB$$

$$= ABC - ACB + ACB - CAB$$

$$= A(BC - CB) + (AC - CA)B$$

$$= A[B, C] + [A, C]B$$

(2) Similar to (1).

#### 6. Differentiation

#### 6.1. The Gateaux Derivative.

**Note 6.1.1.** In this section, we assume all Banach spaces to be over  $\mathbb{R}$ .

**Definition 6.1.2.** Let X, Y be a Banach spaces,  $A \subset X$  open,  $f : A \to Y$ ,  $x_0 \in A$  and  $x \in X$ . Then f is said to be

(1) right-hand-differentiable at  $x_0$  in the direction x if the limit

$$\lim_{t \to 0^+} \frac{f(x_0 + tx) - f(x_0)}{t}$$

exists. If f is right-hand-differentiable at  $x_0$  in the direction x, we define the **right-hand derivative** of f at  $x_0$  in the direction x, denoted by  $d^+f(x_0;x)$ , to be the above limit.

(2) left-hand-differentiable at  $x_0$  in the direction x if the limit

$$\lim_{t \to 0^{-}} \frac{f(x_0 + tx) - f(x_0)}{t}$$

exists. If f is right-hand-differentiable at  $x_0$  in the direction x, we define the **left-hand derivative** of f at  $x_0$  in the direction x, denoted by  $d^-f(x_0;x)$ , to be the above limit.

(3) differentiable at  $x_0$  in the direction x if the limit

$$\lim_{t \to 0} \frac{f(x_0 + tx) - f(x_0)}{t}$$

exists. If f is differentiable at  $x_0$  in the direction x, we define the **derivative** of f at  $x_0$  in the direction x, denoted by  $df(x_0; x)$ , to be the above limit.

**Exercise 6.1.3.** Let X, Y be Banach spaces,  $A \subset X$  open,  $f : A \to \mathbb{R}$  and  $x_0 \in A$ . Then  $df(x_0; 0) = 0$ .

Proof. Clear. 
$$\Box$$

#### Definition 6.1.4. The Gateaux Derivative:

Let X, Y be Banach spaces,  $A \subset X$  open,  $f: A \to Y$  and  $x_0 \in A$ . Then f is said to be

(1) **right-hand Gateaux differentiable** at  $x_0$  if for each  $x \in X$ ,  $d^+f(x_0; x)$  exits. We define the **right-hand Gateaux derivative** of f at  $x_0$ , denoted  $d^+f(x_0): X \to \mathbb{R}$ , to be

$$d^+ f(x_0)(x) = d^+ f(x_0; x)$$

(2) **left-hand Gateaux differentiable** at  $x_0$  if for each  $x \in X$ ,  $d^-f(x_0; x)$  exits. We define the **left-hand Gateaux derivative** of f at  $x_0$ , denoted  $d^-f(x_0): X \to \mathbb{R}$ , to be

$$d^{-}f(x_0)(x) = d^{-}f(x_0; x)$$

(3) Gateaux differentiable at  $x_0$  if for each  $x \in X$ ,  $df(x_0; x)$  exits. We define the Gateaux derivative of f at  $x_0$ , denoted  $df(x_0): X \to \mathbb{R}$ , to be

$$df(x_0)(x) = df(x_0; x)$$

**Definition 6.1.5.** Let Y be a Banach space,  $A \subset \mathbb{R}$  open and  $f: A \to Y$ . Then f is said to be **Gateaux differentiable** if for each  $x \in A$ , f is Gateaux differentiable at x. If f is Gateaux differentiable, we define  $df: A \to Y^X$  by  $x_0 \mapsto df(x_0)$ .

**Exercise 6.1.6.** Let X, Y be Banach spaces,  $A \subset X$  open,  $f, g : A \to Y$ ,  $\lambda \in \mathbb{R}$  and  $x_0 \in A$ . If f, g are Gateaux differentiable at  $x_0$ , then  $f + \lambda g$  is Gateaux differentiable at  $x_0$  and  $d[f + \lambda g](x_0) = df(x_0) + \lambda dg(x_0)$ .

*Proof.* Similar to the case of the derivative from Calc I.

**Exercise 6.1.7.** Let X, Y be Banach spaces,  $A \subset X$  open,  $f : A \to Y$  and  $x_0 \in A$ . Suppose that f is Gateaux differentiable at  $x_0$ . Then for each  $\lambda \in \mathbb{R}$  and  $x \in X$ ,

$$df(x_0)(\lambda x) = \lambda df(x_0)(x)$$

*Proof.* Let  $\lambda \in \mathbb{R}$  and  $x \in X$ . Then

$$df(x_0)(\lambda x) = \lim_{t \to 0} \frac{f(x_0 + t\lambda x) - f(x_0)}{t}$$

$$= \lim_{t \to 0} \lambda \frac{f(x_0 + t\lambda x) - f(x_0)}{\lambda t}$$

$$= \lambda \lim_{t \to 0} \frac{f(x_0 + t\lambda x) - f(x_0)}{\lambda t}$$

$$= \lambda \lim_{t \to 0} \frac{f(x_0 + t\lambda x) - f(x_0)}{t}$$

$$= \lambda df(x_0)(x)$$

**Exercise 6.1.8.** Let X, Y be Banach spaces,  $A \subset X$  open,  $f : A \to Y$ . If f is constant, then f is Gateaux differentiable and for each  $x_0 \in A, x \in X$ ,

$$df(x_0)(x) = 0$$

*Proof.* Suppose that f is constant. Then there exists  $c \in Y$  such that for each  $x \in A$ , f(x) = c. Let  $x_0 \in A, x \in X$ . Then

$$df(x_0)(x) = \lim_{t \to 0} \frac{f(x_0 + tx) - f(x_0)}{t}$$
$$= \lim_{t \to 0} \frac{c - c}{t}$$
$$= 0$$

**Exercise 6.1.9.** Let X, Y be Banach spaces,  $A \subset X$  open,  $f : A \to Y$ . If f is linear, then f is Gateaux differentiable and for each  $x_0 \in A, x \in X$ ,

$$df(x_0)(x) = f(x)$$

*Proof.* Suppose that f is linear. Let  $x_0 \in A, x \in X$ . Then

$$df(x_0)(x) = \lim_{t \to 0} \frac{f(x_0 + tx) - f(x_0)}{t}$$
$$= \lim_{t \to 0} \frac{f(x_0) + tf(x) - f(x_0)}{t}$$
$$= f(x)$$

**Exercise 6.1.10.** There exist Banach spaces X, Y, and  $f: X \to Y$  such that f is Gateaux differentiable and f is nowhere continuous.

**Hint:** use Exercise 6.1.9

Proof. Set  $X = C^1([0,1])$  and Y = C([0,1]). Equip both X and Y with the sup norm. Define  $T: X \to Y$  by Tf = f'. Then Exercise 4.2.2 implies that T is not bounded. Since T is linear, Exercise 6.1.9 implies that T is Gateaux differentiable. Since T is not bounded, Exercise 4.2.5 implies that T is not continuous at 0. Then Exercise 4.2.4 tells us that T is nowhere continuous.

**Exercise 6.1.11.** Set  $A = \{(x, y) \in \mathbb{R}^2 : y = -x^2 \text{ and } x \neq 0\}$ . Define  $f : \mathbb{R}^2 \setminus A \to \mathbb{R}$  by

$$f(x,y) = \begin{cases} 0 & (x,y) = (0,0) \\ \frac{x^4 y}{x^6 + y^3} & \text{otherwise} \end{cases}$$

Then f is Gateaux differentiable at (0,0) and f is not continuous at (0,0).

**Hint:** Consider the set  $B = \{(x, x^2 : x \in \mathbb{R})\} \subset \mathbb{R}^2 \setminus A$ .

**Exercise 6.1.12.** Let Y be a Banach space,  $A \subset \mathbb{R}$  open,  $f : A \to Y$  and  $x_0 \in A$ . Suppose that f is Gateaux differentiable at  $x_0$ . Then  $df(x_0) \in L(\mathbb{R}, Y)$ .

*Proof.* Let  $x, y, \lambda \in \mathbb{R}$ .

(1) The previous exercise implies

$$df(x_0)(x + \lambda y) = df(x_0)((x + \lambda y)1)$$

$$= (x + \lambda y)df(x_0)(1)$$

$$= xdf(x_0)(1) + \lambda ydf(x_0)(1)$$

$$= df(x_0)(x) + \lambda df(x_0)(y)$$

So  $df(x_0): \mathbb{R} \to Y$  is linear.

(2) Since

$$||df(x_0)(x)|| = ||xdf(x_0)(1)||$$
$$= |x|||df(x_0)(1)||$$

We have that  $df(x_0): \mathbb{R} \to Y$  is bounded with  $||df(x_0)|| \le ||df(x_0)(1)||$ .

**Exercise 6.1.13.** Let X be a Banach space,  $A \subset X$  open,  $f : A \to \mathbb{R}$  and  $x_0 \in A$ . If f is Gateaux differentiable at  $x_0$  and f has a local extremum at  $x_0$ , then  $df(x_0) = 0$ .

*Proof.* Suppose that f is Gateaux differentiable at  $x_0$  and f has a local minimum point at  $x_0$ . Then there exists  $\delta > 0$  such that  $B(x_0, \delta) \subset A$  and for each  $y \in B(x_0, \delta)$ ,  $f(x_0) \leq f(y)$ . For the sake of contradiction, suppose that  $df(x_0) \neq 0$ . Then there exists  $x \in X$  such that  $x \neq 0$  and  $df(x_0)(x) \neq 0$ .

First, suppose that  $df(x_0)(x) < 0$ . Choose  $\epsilon = -df(x_0)(x) > 0$ . Then there exists  $t_0 > 0$  such that for each  $t \in B^*(0, t_0)$ ,  $x_0 + tx \in B(x_0, \delta)$  and

$$\left| \frac{f(x_0 + tx) - f(x_0)}{t} - df(x_0)(x) \right| < \epsilon$$

This implies that for each  $t \in B^*(0, t_0)$ ,

$$\frac{f(x_0 + tx) - f(x_0)}{t} < \epsilon + df(x_0)(x)$$

$$= 0$$

and hence  $f(x_0 + tx) < f(x_0)$ , which is a contradiction.

Now, suppose that  $df(x_0)(x) > 0$ . Then

$$df(x_0)(-x) = -df(x_0)(x)$$
  
< 0

Similarly to above, this implies that there exists  $t_0 > 0$  such that for each  $t \in B^*(0, t_0)$ ,  $x_0 - tx \in B(x_0, \delta)$  and  $f(x_0 - tx) < f(x_0)$  which is a contradiction. So  $df(x_0)(x) = 0$  and  $df(x_0) = 0$ .

If f has a local maximum at  $x_0$ , then -f has a local minimum point at  $x_0$ . Then

$$df(x_0) = -d[-f](x_0)$$
$$= -0$$
$$= 0$$

**Exercise 6.1.14.** Let X, Y, Z be a Banach spaces,  $A \subset X$  open,  $B \subset Y$  open,  $f : A \to Y$ ,  $g : B \to Z$  and  $x_0 \in A$ . Suppose that f is affine. If g is Gateaux differentiable at  $f(x_0)$ , then  $g \circ f$  is Gateaux differentiable at  $x_0$  and

$$d(g \circ f)(x_0)(x) = dg(f(x_0))(df(x_0)(x))$$

*Proof.* Suppose that g is Gateaux differentiable at  $f(x_0)$ . Since f is affine, there exists  $h: A \to Y$  and  $c \in Y$  such that h is linear and f = h + c. Then

$$df(x_0) = dh(x_0)$$
$$= h$$

Let  $x \in X$ . Choose  $\delta > 0$  such that for each  $t \in B(0, \delta) \subset \mathbb{R}$ ,  $f(x_0) + th(x) \in B$ . Then for each  $t \in B^*(0, \delta)$ ,

$$g \circ f(x_0 + tx) = g\left(f(x_0) + t\frac{f(x_0 + tx) - f(x_0)}{t}\right)$$
$$= g(f(x_0) + th(x))$$

This implies that

$$d(g \circ f)(x_0) = \lim_{t \to 0} \frac{g \circ f(x_0 + tx) - g(f(x_0))}{t}$$

$$= \lim_{t \to 0} \frac{g(f(x_0) + th(x)) - g(x_0)}{t}$$

$$= dg(f(x_0))(h(x))$$

$$= dg(f(x_0))(df(x_0)(x))$$

#### 6.2. The Frechet Derivative.

**Exercise 6.2.1.** Let X, Y be a normed vector spaces and  $\phi : X \to Y$  linear. If  $\phi(h) = o(\|h\|)$  as  $h \to 0$ , then  $\phi = 0$ .

*Proof.* Let  $h_0 \in X$ . If  $h_0 = 0$ , then  $\phi(h_0) = 0$ . Suppose that  $h_0 \neq 0$ . Define  $(h_n)_{n \in \mathbb{N}} \subset X$  by

$$h_n = \frac{h_0}{n}$$

Then  $h_n \to 0$ . By continuity of  $\phi$  and our initial assumption we have that

$$||h_0||^{-1}\phi(h_0) = \phi\left(\frac{h_0}{||h_0||}\right)$$
$$= \phi\left(\frac{h_n}{||h_n||}\right)$$
$$= \frac{\phi(h_n)}{||h_n||}$$
$$\to 0$$

which implies that  $||h_0||^{-1}\phi(h_0)=0$ . So  $\phi(h_0)=0$  and hence  $\phi=0$ .

**Exercise 6.2.2.** Let X, Y be a normed vector spaces,  $A \subset X$  open,  $f : A \to Y$  and  $x_0 \in A$ . Suppose that there exists  $\phi : X \to Y$  such that  $\phi$  is linear and

$$f(x_0 + h) = f(x_0) + \phi(h) + o(||h||)$$
 as  $h \to 0$ 

then  $\phi$  is unique.

*Proof.* Suppose that there exists  $\psi: X \to Y$  such that  $\psi$  is linear and such that

$$f(x_0 + h) = f(x_0) + \psi(h) + o(||h||)$$
 as  $h \to 0$ 

Then  $\phi(h) - \psi(h) = o(h)$ . Since  $\phi - \psi$  is linear, the previous exercise implies that  $\phi = \psi$ .  $\square$ 

Note 6.2.3. Recall that for Banach spaces X and Y, there isomorphic isometry

$$L(X, L(X, \dots, L(X, Y)) \dots) \to L^n(X, Y)$$

given by  $\phi \mapsto \psi_{\phi}$  where

$$\psi_{\phi}(x_1, x_2, \cdots, x_n) = \phi(x_1)(x_2), \cdots, (x_n)$$

#### Definition 6.2.4. Frechet Derivative:

Let X, Y be a banach spaces,  $A \subset X$  open,  $f: A \to Y$  and  $x_0 \in A$ .

(1) • Then f is said to be **Frechet differentiable at**  $x_0$  if there exists  $Df(x_0) \in L(X,Y)$  such that,

$$f(x_0 + h) = f(x_0) + Df(x_0)(h) + o(||h||)$$
 as  $h \to 0$ 

- If f is Frechet differentiable at  $x_0$ , we define the **Frechet derivative of** f at  $x_0$  to be  $Df(x_0)$ .
- We say that f is Frechet differentiable if for each  $x \in A$ , f is Frechet differentiable at x.
- If f is Frechet differentiable, we define the **Frechet derivative of** f, denoted  $Df: A \to L(X,Y)$ , by  $x \mapsto D^{(1)}f(x)$ .
- (2) Continuing inductively, we set  $D^0f = f$  and for  $n \geq 2$ ,

- f is said to be n-th order Frechet differentiable at  $x_0$  if f is (n-1)-th order Frechet differentiable and  $D^{n-1}f$  is Frechet differentiable at  $x_0$ .
- If f is n-th order Frechet differentiable at  $x_0$ , we define  $D^n f(x_0) \in L^n(X,Y)$  by

$$D^n f(x_0) = D[D^{n-1} f](x_0)$$

- We say that f is n-th order Frechet differentiable if f is (n-1)-th order Frechet differentiable and for each  $x \in A$ ,  $D^{n-1}f$  is Frechet differentiable at x.
- If f is n-th order Frechet differentiable, we define the n-th order Frechet derivative of f, denoted  $D^n f: A \to L^n(X,Y)$  by  $x \mapsto D^n f(x)$
- (3) If f is n-th order differentiable, then f is said to be **continuously** n-th order differentiable if  $D^n f$  is continuous. We define

$$C^n(A,Y) = \{f : A \to Y : f \text{ is continuously } n\text{-th order differentiable}\}$$

**Exercise 6.2.5.** Let X, Y be a banach spaces,  $A \subset X$  open,  $f, g : A \to Y$ ,  $\lambda \in \mathbb{R}$  and  $x_0 \in A$ . If f and g are Frechet differentiable at  $x_0$ , then  $f + \lambda g$  is Frechet differentiable at  $x_0$  and  $D(f + \lambda g)(x_0) = Df(x_0) + \lambda Dg(x_0)$ .

*Proof.* Suppose that f and g are Frechet differentiable at  $x_0$ . Then

$$f(x_0 + h) = f(x_0) + Df(x_0)(h) + o(||h||)$$
 as  $h \to 0$ 

and

$$g(x_0 + h) = g(x_0) + Dg(x_0)(h) + o(||h||)$$
 as  $h \to 0$ 

This implies that

$$(f + \lambda g)(x_0 + h) = f(x_0 + h) + \lambda g(x_0 + h)$$
  
=  $f(x_0) + Df(x_0)(h) + o(||h||) + \lambda g(x_0) + \lambda Dg(x_0)(h) + o(||h||)$   
=  $(f + \lambda g)(x_0) + [Df(x_0) + \lambda Dg(x_0)](h) + o(||h||)$  as  $h \to 0$ 

Since  $Df(x_0) + \lambda Dg(x_0) \in L(X, Y)$ ,  $f + \lambda g$  is Frechet differentiable at  $x_0$  and  $D(f + \lambda g)(x_0) = Df(x_0) + \lambda Dg(x_0)$ .

**Exercise 6.2.6.** Let X, Y be a banach spaces,  $A \subset X$  open,  $f : A \to Y$  and  $x_0 \in A$ . If f is Frechet differentiable at  $x_0$ , then f is continuous at  $x_0$ .

Proof. Suppose that f is Frechet differentiable at  $x_0$ . Then  $f(x) - f(x_0) = Df(x_0)(x - x_0) + o(\|x - x_0\|)$  as  $x \to x_0$ . Hence  $\|f(x) - f(x_0)\| \le \|Df(x_0)\| \|x - x_0\| + o(\|x - x_0\|)$  as  $x \to x_0$ . This implies that  $f(x) \to f(x_0)$  as  $x \to x_0$  and therefore f is continuous at  $x_0$ .

**Exercise 6.2.7.** Let X, Y be a banach spaces,  $A \subset X$  open,  $f : A \to Y$  and  $x_0 \in A$ . If f is Frechet differentiable at  $x_0$ , then f is Gateaux differentiable at  $x_0$  and  $df(x_0) = Df(x_0)$ .

Proof. Suppose that f is Frechet differentiable at  $x_0$ . Then  $f(x_0 + h) = f(x_0) + Df(x_0)(h) + o(||h||)$  as  $h \to 0$ . Let  $x \in X$ . Then  $f(x_0 + tx) - f(x_0) = tDf(x_0)(x) + o(t)$  as  $t \to 0$ . This implies that f is differentiable at  $x_0$  in the direction x and

$$df(x_0)(x) = \lim_{t \to 0} \frac{f(x_0 + tx) - f(x_0)}{t}$$
  
=  $Df(x_0)(x)$ 

Since  $x \in X$  is arbitrary, f is Gateaux differentiable at  $x_0$  and  $df(x_0) = Df(x_0)$ .

**Exercise 6.2.8.** Let X be a Banach space,  $A \subset X$  open,  $f : A \to \mathbb{R}$  and  $x_0 \in A$ . If f is Frechet differentiable at  $x_0$  and f has a local extremum at  $x_0$ , then  $Df(x_0) = 0$ .

*Proof.* Suppose that f is Frechet differentiable at  $x_0$  and f has a local extremum at  $x_0$ . Two previous exercises imply that f is Gateaux differentiable at  $x_0$  and

$$Df(x_0) = df(x_0)$$
$$= 0$$

**Definition 6.2.9.** Let X, Y be a banach spaces,  $A \subset X$  open,  $f : A \to Y$  and  $x_0 \in A$ . Suppose that f is Frechet differentiable at  $x_0$ . Define  $R_f(x_0) : A - x_0 \to Y$  by

$$R_f(x_0)(h) = f(x_0 + h) - f(x_0) - Df(x_0)(h)$$

**Exercise 6.2.10.** Let X, Y be a banach spaces,  $A \subset X$  open,  $f : A \to Y$  and  $x_0 \in A$ . If f is Frechet differentiable at  $x_0$ , then

$$f(x_0 + h) - f(x_0) = O(||h||)$$
 as  $h \to 0$ 

Proof. Suppose that f is Frechet differentiable at  $x_0$ . Then  $R_f(h) = o(\|h\|)$  as  $h \to 0$ . Hence there exists  $\delta > 0$  such that  $B(0, \delta) \subset A - x_0$  and for each  $h \in B(0, \delta)$ ,  $\|R_f(h)\| \le \|h\|$ . Hence for each  $h \in B(0, \delta)$ 

$$||f(x_0 + h) - f(x_0)|| = ||Df(x_0)(h) + R_f(x_0)(h)||$$

$$\leq ||Df(x_0)(h)|| + ||R_f(x_0)(h)||$$

$$\leq ||Df(x_0)|| ||(h)|| + ||h||$$

$$= (||Df(x_0)|| + 1)||h||$$

#### Exercise 6.2.11. Chain Rule:

Let X, Y, Z be a Banach spaces,  $A \subset X$  open,  $B \subset Y$  open,  $f : A \to Y$ ,  $g : B \to Z$  and  $x_0 \in A$ . Suppose that  $f(x_0) \in B$ . If f is Frechet differentiable at  $x_0$  and g is Frechet differentiable at  $f(x_0)$ , then  $g \circ f$  is Frechet differentiable at  $x_0$  and

$$D(g \circ f)(x_0) = Dg(f(x_0)) \circ Df(x_0)$$

*Proof.* Suppose that f is Frechet differentiable at  $x_0$  and g is Frechet differentiable at  $f(x_0)$ .

• The previous exercise implies that there exists  $\delta^* > 0$  and K > 0 such that for each  $h \in B(0, \delta^*)$ ,  $||f(x_0 + h) - f(x_0)|| \le K||h||$ . Let  $\epsilon > 0$ . Since  $R_g(f(x_0))(h') = o(||h'||)$  as  $h' \to 0$ , there exists  $\delta' > 0$  such that for each  $h' \in B(0, \delta')$ ,  $||R_g(f(x_0))(h')|| \le \frac{\epsilon}{K}||h'||$ .

Choose  $\delta = \min(\delta'/K, \delta^*)$ . Let  $h \in B(0, \delta)$ . Then

$$||f(x_0 + h) - f(x_0)|| \le K||h||$$
  
 $< \delta'$ 

This implies that

$$||R_g(f(x_0))(f(x_0+h) - f(x_0))|| \le \frac{\epsilon}{K} ||f(x_0+h) - f(x_0)||$$

$$\le \frac{\epsilon}{K} K ||h||$$

$$\le \epsilon ||h||$$

So  $R_g(f(x_0))(f(x_0+h)-f(x_0))=o(||h||)$  as  $h\to 0$ .

- Since  $||Dg(f(x_0))(R_f(x_0)(h))|| \le ||Dg(f(x_0))|| ||R_f(x_0)(h)||$  and  $R_f(x_0)(h) = o(h)$  as  $h \to 0$ , we have that  $Dg(f(x_0))(R_f(x_0)(h)) = o(h)$  as  $h \to 0$ .
- Combining the previous two observations, we have that  $Dg(f(x_0))(R_f(x_0)(h)) + R_g(f(x_0))(f(x_0+h)-f(x_0)) = o(\|h\|)$  as  $h \to 0$ .
- All together, we obtain

$$g \circ f(x_0 + h) = g(f(x_0)) + f(x_0 + h) - f(x_0))$$

$$= g(f(x_0)) + Dg(f(x_0))(f(x_0 + h) - f(x_0)) + R_g(f(x_0))(f(x_0 + h) - f(x_0))$$

$$= g(f(x_0)) + Dg(f(x_0))(Df(x_0)(h) + R_f(x_0)(h))$$

$$+ R_g(f(x_0))(f(x_0 + h) - f(x_0))$$

$$= g(f(x_0)) + Dg(f(x_0))(Df(x_0)(h)) + Dg(f(x_0))(R_f(x_0)(h))$$

$$+ R_g(f(x_0))(f(x_0 + h) - f(x_0))$$

$$= g \circ f(x_0) + Dg(f(x_0)) \circ Df(x_0)(h) + o(||h||) \text{ as } h \to 0$$

So  $g \circ f$  is Frechet differentiable at  $x_0$  and  $D(g \circ f)(x_0) = Dg(f(x_0)) \circ Df(x_0)$ .

**Exercise 6.2.12.** Let Y be a Banach space,  $A \subset \mathbb{R}$  open and  $f : A \to Y$ . Then f is Gateaux differentiable iff f is Frechet differentiable.

*Proof.* Suppose that f is Gateaux differentiable. Let  $x_0 \in A$ . A previous exercise implies that  $df(x_0) \in L(\mathbb{R}, Y)$ . By defintion,

$$\lim_{h \to 0} \left\| \frac{f(x_0 + h) - f(x_0)}{h} - df(x_0)(1) \right\| = 0$$

This is equivalent to saying that

$$f(x_0 + h) = f(x_0) + df(x_0)(h) + o(|h|)$$
 as  $h \to 0$ 

So f is Frechet differentiable at  $x_0$  and  $Df(x_0) = df(x_0)$ .

## 6.3. The Calc I Derivative.

## Definition 6.3.1. Calc I Derivative:

Let Y be a Banach space,  $A \subset \mathbb{R}$  open,  $f: A \to Y$  and  $x_0 \in A$ .

(1) • If f is Frechet differentiable at  $x_0$ , we define the **calc I derivative of** f **at**  $x_0$ , denoted

$$f'(x_0)$$
 or  $\frac{\mathrm{d}f}{\mathrm{d}t}(x_0)$ 

by

$$f'(x_0) = \lim_{t \to 0} \frac{f(x_0 + t) - f(x_0)}{t}$$
$$= df(x_0)(1)$$
$$= Df(x_0)(1)$$

- If f is Frechet differentiable, we define  $f': A \to Y$  by  $x \mapsto f'(x)$ .
- (2) Continuing inductively, we set  $f^{(0)} = f$  and for  $n \ge 1$ ,
  - if  $f^{(n-1)}$  is Frechet differentiable at  $x_0$ , we define the (n)-th order calc I derivative of f at  $x_0$ , denoted  $f^{(n)}(x_0)$ , by

$$f^{(n)}(x_0) = [f^{(n-1)}]'(x_0)$$

• if  $f^{(n-1)}$  is Frechet differentiable, we define  $f^{(n)}: A \to Y$  by

$$f^{(n)} = [f^{(n-1)}]'$$

**Exercise 6.3.2.** Let Y be a Banach space,  $A \subset \mathbb{R}$  open and  $f : A \to Y$ . If f is n-th order Frechet differentiable, then for each  $x_0 \in A$  and  $k \in \{1, \dots, n\}$ ,

$$f^{(k)}(x_0) = D^k f(x_0)(1^{\oplus k})$$

*Proof.* Let  $x_0 \in A$ . We proceed by induction. The base case is true by definition. Let  $k \in \{1, \dots, n\}$ . Suppose the claim is true for k - 1. Then

$$f^{(k-1)}(x_0) = D^{k-1}f(x_0)(1^{\oplus (k-1)})$$

Since f is n-th order Frechet differentiable,

$$D^{k-1}f(x_0+h) = D^{k-1}f(x_0) + D^kf(x_0)(h) + o(||h||)$$
 as  $h \to 0$ 

This implies that

$$f^{(k-1)}(x_0 + h) = D^{k-1}f(x_0 + h)(1^{\oplus (k-1)})$$
  
=  $D^{k-1}f(x_0)(1^{\oplus (k-1)}) + D^kf(x_0)(h)(1^{\oplus (k-1)}) + o(||h||)$  as  $h \to 0$ 

Therefore for each  $h \in \mathbb{R}$ ,

$$Df^{(k-1)}(x_0)(h) = D^k f(x_0)(h)(1^{\oplus (k-1)})$$

and by definition,

$$f^{(k)}(x_0) = [f^{(k-1)}]'(x_0)$$
$$= Df^{(k-1)}(x_0)(1)$$
$$= D^k f(x_0)(1^{\oplus k})$$

**Exercise 6.3.3.** Let X, Y be Banach spaces,  $A \subset X$  open,  $f \in C^n(A, Y), x_0 \in A$ , and  $h \in X$ . Suppose that  $\{x_0 + th : t \in [0, 1]\} \subset A$ . Define and  $g : (0, 1) \to Y$  by

$$g(t) = f(x_0 + th)$$

Then for each  $k \in \{1..., n\}$  and  $t \in (0, 1)$ ,

$$g^{(k)}(t) = D^k f(x_0 + th)(h^{\oplus k})$$

*Proof.* We proceed by induction. It is straightforward to show that the claim is true for k = 1.

Let 
$$k \in \{1..., n\}$$
. Suppose that  $g^{(k-1)}(t) = D^{k-1}f(x_0 + th)(h^{\oplus (k-1)})$ . Since  $f \in C^k(A, Y)$ ,  $D^{k-1}f(x_0 + s_0h + th) = D^{k-1}f(x_0 + s_0h) + D^kf(x_0 + s_0h)(th) + o(||t||)$  as  $t \to 0$ 

The previous exercise implies that

$$g^{(k-1)}(s_0 + t) = D^{k-1}g(s_0 + t)(1^{\oplus (k-1)})$$

$$= D^{k-1}f(x_0 + s_0h + th)(h^{\oplus (k-1)})$$

$$= D^{k-1}f(x_0 + s_0h)(h^{\oplus (k-1)}) + D^kf(x_0 + s_0h)(th)(h^{\oplus (k-1)}) + o(||t||) \text{ as } t \to 0$$

Hence

$$Dg^{(k-1)}(s_0)(t) = D^k f(x_0 + s_0 h)(th)(h^{\oplus (k-1)})$$

and

$$g^{(k)}(t) = Dg^{(k-1)}(t)(1)$$
  
=  $D^k f(x_0 + th)(h^{\oplus k})$ 

### 6.4. Mean Value Theorem.

**Exercise 6.4.1.** Let X be a Banach space,  $A \subset X$  open and convex, and  $f: A \to \mathbb{R}$ . If f is continuous and Gateaux differentiable, then for each  $x, y \in A$ , there exists  $t^* \in (0,1)$  such that  $f(x) - f(y) = df(t^*x + (1-t^*)y)(x-y)$ .

Proof. Suppose that f is continuous and Gateaux differentiable. Let  $x, y \in A$ . The claim is clearly true when f(x) = f(y). Suppose that  $f(x) \neq f(y)$ . Define  $h : [0,1] \to X$  by h(t) = tx + (1-t)y. Set  $g = f \circ h : [0,1] \to \mathbb{R}$ . Then g is continuous on [0,1] and Exercise 6.1.14 implies that g is Gateaux differentiable on (0,1). Then Exercise 6.2.12 Exercise 6.1.14 and the mean value theorem implies that there exists  $t^* \in (0,1)$  such that

$$f(x) - f(y) = g(1) - g(0)$$

$$= g'(t^*)$$

$$= dg(t^*)(1)$$

$$= df(h(t^*))(dh(t^*)(1))$$

$$= df(h(t^*))(h'(t^*))$$

$$= df(t^*x + (1 - t^*)y)(x - y)$$

**Exercise 6.4.2.** Let X be a Banach space,  $A \subset X$  open and convex, and  $f: A \to \mathbb{R}$ . If f is Frechet differentiable, then for each  $x, y \in A$ , there exists  $t^* \in (0,1)$  such that  $f(x) - f(y) = Df(t^*x + (1-t^*)y)(x-y)$ .

*Proof.* Suppose that f is Frechet differentiable. Then f is continuous and Gateaux differentiable. Now apply the previous exercise.

#### Exercise 6.4.3. Mean Value Theorem:

Let X, Y be a Banach spaces,  $A \subset X$  open and convex and  $f : A \to Y$ . If f is Frechet differentiable, then for each  $x, y \in A$ , there exists  $t^* \in (0,1)$  such that

$$||f(x) - f(y)|| \le ||Df(t^*x + (1 - t^*)y)||||x - y||$$

**Hint:** For  $x, y \in A$  with  $f(x) \neq f(y)$ , using a Hahn-Banach argument, find  $\lambda \in Y^*$  such that  $\|\lambda\| = 1$  and  $\lambda(f(x) - f(y)) = \|f(x) - f(y)\|$ .

*Proof.* Suppose that f is Frechet differentiable. Let  $x, y \in A$ . The claim is clearly true when f(x) = f(y). Suppose that  $f(x) \neq f(y)$ . An exercise in the section on linear functionals implies that there exists  $\lambda \in Y^*$  such that  $\lambda(f(x) - f(y)) = ||f(x) - f(y)||$  and  $||\lambda|| = 1$  Define  $g: [0,1] \to \mathbb{R}$  by

$$g(t) = \lambda(f(tx + (1 - t)y))$$

Then g is continuous and (Frechet) differentiable on (0,1) with

$$Dg(t)(h) = \lambda \circ Df(tx + (1-t)y)((x-y)h)$$

which implies that

$$g'(t) = Dg(t)(1)$$
  
=  $\lambda \circ Df(tx + (1-t)y)((x-y))$ 

The mean value theorem implies that there exists  $t^* \in (0,1)$  such that

$$||f(x) - f(y)|| = \lambda(f(x) - f(y))$$

$$= g(1) - g(0)$$

$$= g'(t^*)$$

$$= \lambda \circ Df(t^*x + (1 - t^*)y)((x - y))$$

Taking absolute values, we see that

$$||f(x) - f(y)|| = |\lambda \circ Df(t^*x + (1 - t^*)y)((x - y))|$$

$$\leq ||\lambda|| ||Df(t^*x + (1 - t^*)y)|| ||x - y||$$

$$\leq ||Df(t^*x + (1 - t^*)y)|| ||x - y||$$

**Exercise 6.4.4.** Let X, Y be a Banach spaces,  $A \subset X$  open and convex and  $f : A \to Y$ . Suppose that f is Frechet differentiable. If for each  $x \in A$ , Df(x) = 0, then f is constant.

*Proof.* Suppose that for each  $x \in A$ , Df(x) = 0. Let  $x, y \in A$ . Then the mean value theorem implies that there exists  $t \in (0,1)$  such that

$$||f(x) - f(y)|| \le ||Df(tx + (1 - t)y)|| ||x - y||$$

$$= 0$$

So 
$$f(x) = f(y)$$
.

**Exercise 6.4.5.** Let X, Y be a Banach spaces,  $A \subset X$  open and convex and  $f, g : A \to Y$ . Suppose that f and g are Frechet differentiable. If Df = Dg, then there exists  $c \in Y$  such that f = g + c.

*Proof.* Suppose that Df = Dg. Then D(f - g) = 0 and the previous exercise implies that f - g is constant.

**Exercise 6.4.6.** Let X, Y be a Banach spaces,  $A \subset \mathbb{R}$  open and  $f : A \to Y$ . Suppose that f is Frechet differentiable. Then  $f' \in C(A, Y)$  iff  $f \in C^1(A, Y)$ .

*Proof.* Suppose that  $f' \in C(A, Y)$ . Let  $x, y \in A$  and  $h \in \mathbb{R}$ . Then

$$||(Df(x) - Df(y))(h)|| = ||Df(x)(h) - Df(y)(h)||$$

$$= ||hf'(x) - hf'(y)||$$

$$= ||h(f'(x) - f'(y))||$$

$$= ||f'(x) - f'(y)||h|$$

So  $||Df(x) - Df(y)|| \le ||f'(x) - f'(y)||$ . Hence continuity of f' implies continuity of Df and  $f \in C^1(A, Y)$ . Conversely, suppose that  $f \in C^1(A, Y)$ . Let  $x, y \in A$ . Then

$$||f'(x) - f'(y)|| = ||Df(x)(1) - Df(y)(1)||$$
$$= ||(Df(x) - Df(y))(1)||$$
$$\le ||Df(x) - Df(y)||$$

Hence continuity of Df implies continuity of f' and  $f' \in C(A, Y)$ .

### 6.5. Taylor's Theorem.

Note 6.5.1. This section makes use of the Bochner integral. For reference, see .

**Exercise 6.5.2.** Let Y be a separable Banach space,  $f:[a,b] \to Y$  continuous so that f is Bochner-integrable. Define  $F:(a,b) \to Y$  by

$$F(x) = \int_{(a,x]} f dm$$

Then  $F \in C^1((a,b),Y)$  and for each  $x_0 \in (a,b)$  and  $F'(x_0) = f(x_0)$ .

*Proof.* Let  $x_0 \in (a, b)$  and  $h \in (0, b - x_0)$ . Then continuity implies that

$$\frac{1}{\|h\|} \left| \int_{(x_0, x_0 + h]} f - f(x_0) dm \right| \leq \frac{1}{\|h\|} \max_{x \in (x_0, x_0 + h]} |f(x) - f(x_0)| \|h\|$$

$$= \max_{x \in [x_0, x_0 + h]} |f(x) - f(x_0)|$$

$$\to 0 \text{ as } h \to 0$$

So

$$\int_{(x_0, x_0 + h]} f - f(x_0) dm = o(\|h\|) \quad \text{as } h \to 0$$

Therefore

$$F(x_0 + h) = \int_{(a,x_0+h]} fdm$$

$$= \int_{(a,x_0]} fdm + \int_{(x_0,x_0+h]} fdm$$

$$= \int_{(a,x_0]} fdm + hf(x_0) + \int_{(x_0,x_0+h]} f - f(x_0)dm$$

$$= F(x_0) + hf(x_0) + o(||h||) \quad \text{as } h \to 0$$

The case is similar for  $h \in (x_0 - b, 0)$ . Since the map  $h \mapsto f(x_0)h$  is bounded, F is Frechet differentiable at  $x_0$  and  $DF(x_0)(h) = f(x_0)h$ . This implies that  $F'(x_0) = f(x_0)$  and the previous exercise implies tells us that continuity of f implies continuity of DF. So  $F \in C^1(A, Y)$ .

**Exercise 6.5.3. Fundamental Theorem of Calculus:** Let Y be a separable Banach space and  $f \in C^1((a,b),Y)$ . Then for each  $x, x_0 \in (a,b), x_0 < x$  implies that

- (1) f' is Bochner integrable on  $(x_0, x]$
- (2)

$$f(x) - f(x_0) = \int_{(x_0, x]} f' dm$$

Proof. (1) Since  $f \in C^1((a,b),Y)$ , a previous exercise tells us that  $f' \in C_Y(a,b)$ . Let  $x, x_0 \in (a,b)$ . Suppose that  $x_0 < x$ . Choose  $c, d \in (a,b)$  such that  $a < c < x_0 < x < d < b$ . Then f' is continuous on [c,d] and hence Bochner-integrable on (c,d] and  $(x_0,x]$ .

(2) Define  $g:(c,d)\to Y$  by

$$g(\xi) = \int_{(c,\xi]} f' dm$$

Then the previous exercise implies that  $g \in C_Y^1(c,d)$  and for each  $t \in (c,d)$ , g'(t) = f'(t). Let  $t \in (c,d)$  and  $h \in \mathbb{R}$ . Then

$$Dg(t)(h) = hg'(t)$$
$$= hf'(t)$$
$$= Df(t)(h)$$

So Dg = Df on (c, d). A previous exercise implies that there exists  $c \in Y$  such that f = g + c on (c, d). Then

$$f(x) - f(x_0) = g(x) + c - (g(x_0) + c)$$

$$= g(x) - g(x_0)$$

$$= \int_{(c,x]} f' dm - \int_{(c,x_0]} f' dm$$

$$= \int_{(x_0,x]} f' dm$$

**Exercise 6.5.4.** Let Y be a Banach space,  $A \subset \mathbb{R}$  open and  $g: A \to Y$ . If g is n-th order Frechet differentiable, then

$$\frac{\mathrm{d}}{\mathrm{d}t} \sum_{k=0}^{n-1} \frac{(1-t)^k}{k!} g^{(k)}(t) = \frac{(1-t)^{n-1}}{(n-1)!} g^{(n)}(t)$$

*Proof.* Taking the derivative yields a telescoping series.

#### Exercise 6.5.5. Taylor's Theorem I:

Let X be a Banach space, Y a separable Banach space,  $A \subset X$  open and convex,  $f \in C^{n+1}(A,Y)$ ,  $x_0 \in A$ , and  $h \in A - x_0$ . Then

$$f(x_0 + h) = \sum_{k=0}^{n} \frac{1}{k!} D^k f(x_0)(h^{\oplus k}) + R(x_0; h)$$

where  $R(x_0; \cdot): A - x_0 \to Y$  is defined by

$$R(x_0; h) = \frac{1}{n!} \int_{(0,1)} (1-t)^n D^{n+1} f(x_0 + th) (h^{\oplus (n+1)}) dm(t)$$

and  $R(x_0; h) = o(||h||^n)$  as  $h \to 0$ .

**Hint:** Define  $g:(0,1)\to Y$  by

$$g(t) = f(x_0 + th)$$

Then use the previous exercise and the fundamental theorem of calculus.

*Proof.* Let  $h \in X$ . Suppose that  $x_0 + h \in A$ . Define  $g:(0,1) \to Y$  by

$$g(t) = f(x_0 + th)$$

For each  $k \in \{1, ..., n+1\}$ , a previous exercise implies that  $g^{(k)}(t) = D^k f(x_0 + th)(h^{\oplus k})$ , so  $g^{(k)}(0) = D^k f(x_0)(h^{\oplus k})$ . The previous exercise and the fundamental theorem of calculus tell us that

$$f(x_0 + h) - \sum_{k=0}^{n} \frac{1}{k!} D^k f(x_0)(h^{\oplus k}) = g(1) - \sum_{k=0}^{n} \frac{1}{k!} g^{(k)}(0)$$

$$= \int_{(0,1)} \left[ \frac{\mathrm{d}}{\mathrm{d}t} \sum_{k=0}^{n} \frac{(1-t)^k}{k!} g^{(k)}(t) \right] dm(t)$$

$$= \int_{(0,1)} \frac{(1-t)^n}{n!} g^{(n+1)}(t) dm(t)$$

$$= \frac{1}{n!} \int_{(0,1)} (1-t)^n D^{n+1} f(x_0 + th)(h^{\oplus (n+1)}) dm(t)$$

$$= R(x_0; h)$$

Note that

$$\frac{1}{n+1} = \frac{1}{n!} \int_{(0,1)} (1-t)^n dm(t)$$

Since  $D^{n+1}f$  is continuous at  $x_0$ , there exists  $\delta_1 > 0$  such that for each  $h \in B(0, \delta_1)$ ,  $x_0 + h \in A$  and

$$||D^{n+1}f(x_0+h) - D^{n+1}f(x_0)|| < 1$$

Let  $\epsilon > 0$ . Choose  $\delta_2 > 0$  such that

$$\frac{1}{n+1} \left( \|D^{n+1} f(x_0)\| + 1 \right) \delta_2 < \epsilon$$

Set  $\delta = \min(\delta_1, \delta_2)$ . Let  $h \in B(0, \delta)$ . Then

$$||R(x_0;h)|| = \left\| \int_{(0,1)} \frac{1}{n!} \int_{(0,1)} (1-t)^n D^{n+1} f(x_0 + th) (h^{\oplus (n+1)}) dm(t) \right\|$$

$$\leq \frac{1}{n!} \int_{(0,1)} ||(1-t)^n D^{n+1} f(x_0 + th) (h^{\oplus (n+1)})|| dm(t)$$

$$\leq \max_{t \in [0,1]} ||D^{n+1} f(x_0 + th)|| ||h||^{n+1} \frac{1}{n!} \int_{(0,1)} (1-t)^n dm(t)$$

$$\leq \frac{1}{n+1} \left( ||D^{n+1} f(x_0)|| + \max_{t \in [0,1]} ||D^{n+1} f(x_0 + th) - D^{n+1} f(x_0)|| \right) ||h||^{n+1}$$

$$< \frac{1}{n+1} \left( ||D^{n+1} f(x_0)|| + 1 \right) ||h||^{n+1}$$

$$< \epsilon ||h||^n$$

So 
$$R(x_0, h) = o(||h||^n)$$
 as  $h \to 0$ .

### Exercise 6.5.6. Taylor's Theorem II:

Let X be a Banach space, Y a separable Banach space,  $A \subset X$  open and convex,  $f \in$ 

 $C^n(A,Y), x_0 \in A$ , and  $h \in A - x_0$ . Then there exists  $R_f(x_0): A - x_0 \to Y$  such that

$$f(x_0 + h) = \sum_{k=0}^{n} \frac{1}{k!} D^k f(x_0)(h^{\oplus k}) + R_f(x_0)(h)$$

and  $R_f(x_0)(h) = o(||h||^n)$  as  $h \to 0$ .

Hint: use Taylor's theorem and expand the derivative inside the integral.

*Proof.* This is clear by definition for n = 1. Suppose that  $n \ge 2$ . Taylor's theorem implies that

$$f(x_0 + h) = \sum_{k=0}^{n-2} \frac{1}{k!} D^k f(x_0)(h^{\oplus k}) + S(x_0; h)$$

where  $S(x_0; \cdot): A - x_0 \to Y$  is defined by

$$S(x_0; h) = \frac{1}{(n-2)!} \int_{(0,1)} (1-t)^{n-2} D^{n-1} f(x_0 + th) (h^{\oplus (n-1)}) dm(t)$$

and  $S(x_0;h) = o(\|h\|^n)$  as  $h \to 0$ . Define  $T_f^{n-1}(x_0): A - x_0 \to L^{n-1}(X;Y)$  by

$$T_f^{n-1}(x_0)(h) = D^{n-1}f(x_0+h) - D^{n-1}f(x_0) - D^nf(x_0)(h)$$

so that

$$D^{n-1}f(x_0+h) = D^{n-1}f(x_0) + D^nf(x_0)(h) + T_f^{n-1}(x_0)(h)$$

and  $T_f^{n-1}(x_0)(h) = o(||h||)$  as  $h \to 0$ . Define  $R_f(x_0) : A - x_0 \to Y$  by

$$R_f(x_0)(h) = \frac{1}{(n-2)!} \int_{(0,1)} (1-t)^{n-2} T_f^{n-1}(x_0)(th) (h^{\oplus (n-1)}) dm(t)$$

Note that

 $\int_0^1 (1-t)^{n-2} dt = \frac{1}{n-1}$ 

 $\int_0^1 (1-t)^{n-2}tdt = \frac{1}{n(n-1)}$ 

Let  $\epsilon > 0$ . Choose  $\delta > 0$  such that for each  $h \in B(0, \delta)$ ,  $h \in A - x_0$  and

$$||T_f^{n-1}(x_0)(h)|| \le \epsilon n! ||h||$$

Let  $h \in B(0, \delta)$ . Then

$$||R_{f}(x_{0})(h)|| = \left\| \frac{1}{(n-2)!} \int_{(0,1)} (1-t)^{n-2} T_{f}^{n-1}(x_{0})(th)(h^{\oplus(n-1)}) dm(t) \right\|$$

$$\leq \frac{1}{(n-2)!} \int_{(0,1)} (1-t)^{n-2} ||T_{f}^{n-1}(x_{0})(th)(h^{\oplus(n-1)})|| dm(t)$$

$$\leq \frac{1}{(n-2)!} \int_{(0,1)} (1-t)^{n-2} ||T_{f}^{n-1}(x_{0})(th)|| ||h||^{n-1} dm(t)$$

$$\leq \frac{\epsilon}{(n-2)!} n! ||h||^{n} \int_{(0,1)} (1-t)^{n-2} t dm(t)$$

$$= \epsilon ||h||^{n}$$

So that  $R_f(x_0)(h)(h^{\oplus (n-1)}) = o(||h||^n)$  as  $h \to 0$ .

$$S(x_0; h) = \frac{1}{(n-2)!} \int_{(0,1)} (1-t)^{n-2} D^{n-1} f(x_0 + th) (h^{\oplus (n-1)}) dm(t)$$

$$= \frac{1}{(n-2)!} \int_{(0,1)} (1-t)^{n-2} D^{n-1} f(x_0) (h^{\oplus (n-1)}) dm(t)$$

$$+ \frac{1}{(n-2)!} \int_{(0,1)} (1-t)^{n-2} t D^n f(x_0) (h) (h^{\oplus (n-1)}) dm(t)$$

$$+ \frac{1}{(n-2)!} \int_{(0,1)} (1-t)^{n-2} T_f^{n-1}(x_0) (th) (h^{\oplus (n-1)}) dm(t)$$

$$= \frac{1}{(n-1)!} D^{n-1} f(x_0) (h^{\oplus (n-1)}) + \frac{1}{n!} D^n f(x_0) (h^{\oplus n}) + R_f(x_0) (h)$$

Hence

$$f(x_0 + h) = \sum_{k=0}^{n-2} \frac{1}{k!} D^k f(x_0) (h^{\oplus k}) + S(x_0; h)$$
$$= \sum_{k=0}^n \frac{1}{k!} D^k f(x_0) (h^{\oplus k}) + R_f(x_0) (h)$$

**Exercise 6.5.7.** Let X be a Banach space,  $A \subset X$  open and convex and  $f \in C^2(A)$ ,  $x_0 \in A$ . If f has a local minimum at  $x_0$ , then  $D^2 f(x_0)$  is positive semidefinite.

*Proof.* Suppose that f has a local minimum at  $x_0$ , then  $Df(x_0) = 0$ . Let  $x \in X$ . Then

$$0 \le f(x+h) - f(x_0)$$
  
=  $\frac{1}{2}D^2 f(x_0)(h,h) + o(\|h\|^2)$  as  $h \to 0$ 

Let  $h \in X$ . Then

$$0 \le \frac{1}{2}t^2D^2f(x_0)(h,h) + o(t^2)$$
 as  $t \to 0$ 

This implies that  $D^2 f(x_0)(h,h) \ge 0$ . So  $D^2 f(x_0)$  is positive semidefinite.

# 6.6. The Gradient.

**Definition 6.6.1.** Let H be a Hilbert space,  $f: H \to \mathbb{R}$  and  $x_0 \in H$ . Suppose that f is Frechet differentiable at  $x_0$ . Then  $Df(x_0) \in H^*$ . We define the **gradient of** f **at**  $x_0$ , denoted  $\nabla f(x_0) \in H$ , via the Riesz representation theorem to be the unique element of H satisfying

$$\langle \nabla f(x_0), y \rangle = Df(x_0)(y)$$
 for each  $y \in H$ 

#### 7. Convexity

### 7.1. Introduction.

Note 7.1.1. In this section, we assume all vector spaces are real.

**Definition 7.1.2.** Let X be a vector space and  $A \subset X$ . Then A is said to be **convex** if for each  $x, y \in A$ , and  $t \in [0, 1]$ ,  $tx + (1 - t)y \in A$ .

**Definition 7.1.3.** Let X be a vector space and  $f: A \to R$ . Then f is said to be **convex** if for each  $x, y \in A$ ,  $t \in [0, 1]$ ,

$$f(tx + (1-t)y) \le tf(x) + (1-t)f(y)$$

**Definition 7.1.4.** Let X be a vector space and  $f: A \to R$ . Then f is said to be **strictly convex** if for each  $x, y \in A$ ,  $t \in (0,1)$ ,  $x \neq y$  implies that

$$f(tx + (1-t)y) < tf(x) + (1-t)f(y)$$

**Exercise 7.1.5.** Let X be a vector space,  $f \in X^*$  and  $g : X \to \mathbb{R}$  constant. Then f and g are convex.

*Proof.* Let  $x, y \in X$  and  $t \in [0, 1]$ . Put c = g(0). Then

$$f(tx + (1 - t)y) = tf(x) + (1 - t)f(y)$$

and

$$g(tx + (1-t)y) = c$$

$$= tc + (1-t)c$$

$$= tq(x) + (1-t)q(y)$$

So f and q are convex.

**Exercise 7.1.6.** Let X be a vector space,  $A \subset X$  convex,  $f, g : A \to \mathbb{R}$  and  $\lambda \geq 0$ . If f, g are convex, then

- (1) f + g is convex
- (2)  $\lambda f$  is convex

*Proof.* Suppose that f and g are convex. Let  $x, y \in A$  and  $t \in [0, 1]$ . Then

$$(f + \lambda g)(tx + (1 - t)y) = f(tx + (1 - t)y) + \lambda g(tx + (1 - t)y)$$

$$\leq tf(x) + (1 - t)f(y) + t\lambda g(x) + (1 - t)\lambda g(y)$$

$$= t(f(x) + \lambda g(x)) + (1 - t)(f(y) + \lambda g(y))$$

$$= t(f + \lambda g)(x) + (1 - t)(f + \lambda g)(y)$$

**Definition 7.1.7.** Let X be a vector space and  $f: X \to \mathbb{R}$ . Then f is said to be **affine** if there exists  $\phi \in X^*$ ,  $a \in R$  constant such that  $f = \phi + a$ .

**Exercise 7.1.8.** Let X be a vector space and  $f: X \to \mathbb{R}$ . If f is affine, then f is convex.

*Proof.* Suppose that f is affine. Then there exists  $\phi \in X^*$ ,  $a \in R$  constant such that  $f = \phi + a$ . Then  $\phi$  is convex and  $g: X \to \mathbb{R}$  defined by g(x) = a is convex. So  $f = \phi + g$  is convex.

**Exercise 7.1.9.** Let X be a vector space,  $A \subset X$  convex,  $f : \mathbb{R} \to \mathbb{R}$  and  $g : A \to \mathbb{R}$ . If f is convex and increasing and g is convex, then  $f \circ g$  is convex.

*Proof.* Let  $t \in [0,1]$  and  $x,y \in A$ . Then convexity of g implies that

$$g(tx + (1-t)y) \le tg(x) + (1-t)g(y)$$

and we have

$$f \circ g(tx + (1 - t)y) = f(g(tx + (1 - t)y))$$

$$\leq f(tg(x) + (1 - t)g(y)) \qquad (f \text{ increasing})$$

$$\leq tf(g(x)) + (1 - t)f(g(y)) \qquad (f \text{ convex})$$

$$= tf \circ g(x) + (1 - t)f \circ g(y)$$

So  $f \circ g$  is convex.

**Exercise 7.1.10.** Let X be a vector space,  $A \subset X$  convex,  $f : A \to \mathbb{R}$  convex and  $x_0 \in A$ . Then f has a local minimum point at  $x_0$  iff f has a global minimum point at  $x_0$ .

Proof. If f has a global minimum point at  $x_0$ , then f has a local minimum point at  $x_0$ . Conversely, suppose that f has a local minimum point at  $x_0$ . Then there exists  $\delta > 0$  such that for each  $x \in B(x_0, \delta) \cap A$ ,  $f(x_0) \leq f(x)$ . For the sake of contradiction, suppose that f does not have a global minimum point at  $x_0$ . Then there exists  $x' \in A$  such that  $f(x') < f(x_0)$ . Put  $t_0 = \min(\frac{\delta}{\|x' - x_0\| + 1}, 1) > 0$ . Let  $t \in (0, t_0)$ , then

$$||(tx' + (1 - t)x_0) - x_0|| = t||x' - x_0||$$

$$< \frac{||x' - x_0||\delta}{||x' - x_0|| + 1}$$

$$< \delta$$

so that  $tx' + (1-t)x_0 \in B(x_0, \delta) \cap A$  and hence  $f(x_0) \leq f(tx' + (1-t)x_0)$ . Therefore

$$f(x_0) \le f(tx' + (1-t)x_0)$$
  
 $\le tf(x') + (1-t)f(x_0)$  (convexity of  $f$ )  
 $< tf(x_0) + (1-t)f(x_0)$   
 $= f(x_0)$ 

which is a contradiction. Hence f has a global minimum point at  $x_0$ .

**Exercise 7.1.11.** Let X be a vector space,  $A \subset X$  convex,  $f : A \to \mathbb{R}$  strictly convex. Then f has at most one local minimum point.

*Proof.* For the sake of contradiction suppose that f has at least two local minimum points. Choose  $x_0, x_1 \in A$  such that  $x_0 \neq x_1$  and f has a local minimum point at  $x_0$  and  $x_1$ . The previous exercise implies that f has a global minimum point at  $x_0$  and  $x_1$ . This implies  $f(x_0) = f(x_1)$ . Set t = 1/2. Strict convexity implies that

$$f(tx_0 + (1-t)x_1) < tf(x_0) + (1-t)f(x_1)$$
  
=  $f(x_0)$ 

which is a contradiction since f has a global minimum point at  $x_0$ .

**Definition 7.1.12.** Let X, Y be vector spaces,  $A \subset X \oplus Y$ . For  $y \in Y$ , define

$$A^y = \{ x \in X : (x, y) \in A \}$$

and  $f^y: A^y \to \mathbb{R}$  by

$$f^y(x) = f(x, y)$$

**Exercise 7.1.13.** Let X, Y be vector spaces,  $A \subset X \oplus Y$  convex and  $f : A \to \mathbb{R}$  convex. Then for each  $y \in \pi_2(A)$ ,

- (1)  $A^y$  is convex
- (2)  $f^y$  is convex

where  $\pi_2: X \times Y \to Y$ , the canonical projection of  $X \times Y$  onto Y given by  $\pi_2(x,y) = y$ .

*Proof.* Let  $y \in \pi_2(A)$ ,  $x_1, x_2 \in A^y$  and  $t \in [0, 1]$ . Then by definition,  $(x_1, y)$ ,  $(x_2, y) \in A$ .

- (1) Convexity of A implies that  $(tx_1 + (1-t)x_2, y) \in A$ . Hence  $tx_1 + (1-t)x_2 \in A^y$  and  $A^y$  is convex.
- (2) Convexity of f implies that

$$f^{y}(tx_{1} + (1-t)x_{2}) = f(tx_{1} + (1-t)x_{2}, y)$$

$$= f(t(x_{1}, y) + (1-t)(x_{2}, y))$$

$$\leq tf(x_{1}, y) + (1-t)f(x_{2}, y)$$

$$= tf^{y}(x_{1}) + (t-t)f^{y}(x_{2})$$

and so  $f^y$  is convex.

**Exercise 7.1.14.** Let X, Y be vector spaces and  $A \subset X, B \subset Y$ . If A and B are convex, then  $A \times B \subset X \oplus Y$  is convex.

*Proof.* Suppose that A and B are convex. Let  $(x_1, y_1), (x_2, y_2) \in A \times B$  and  $t \in [0, 1]$ . Convexity of A and B implies that  $tx_1 + (1 - t)x_2 \in A$  and  $ty_1 + (1 - t)y_2 \in B$ . Therefore

$$t(x_1, y_1) + (1 - t)(x_2, y_2) = (tx_1 + (1 - t)x_2, ty_1 + (1 - t)y_2)$$
  
 $\in A \times B$ 

**Exercise 7.1.15.** Let X, Y be vector spaces and  $A \subset X$ ,  $B \subset Y$  convex (implying that  $A \times B$  is convex) and  $f: A \times B \to \mathbb{R}$  convex. Suppose that for each  $y \in B$ ,  $\{f(x,y): x \in A\}$  is bounded below. Then  $\inf_{u \in B} f^y$  is convex

*Proof.* Put  $g = \inf_{y \in B} f^y$ . Let  $x_1, x_2 \in A$ ,  $y_1, y_2 \in B$  and  $t \in [0, 1]$ . Put  $y' = ty_1 + (1 - t)y_2$ . Then convexity of f implies that

$$g(tx_1 + (1-t)x_2) \le f^{y'}(tx_1 + (1-t)x_2)$$

$$= f(tx_1 + (1-t)x_2, ty_1 + (1-t)y_2)$$

$$= f(t(x_1, y_1) + (1-t)(x_2, y_2))$$

$$\le tf(x_1, y_1) + (1-t)f(x_2, y_2)$$

$$= tf^{y_1}(x_1) + (1-t)f^{y_2}(x_2)$$

Since  $y_1 \in B$  is arbitrary, we have that

$$g(tx_1 + (1-t)x_2) \le tg(x_1) + (1-t)f^{y_2}(x_2)$$

Similarly, since  $y_2 \in B$  is arbitrary, we have that

$$g(tx_1 + (1-t)x_2) \le tg(x_1) + (1-t)g(x_2)$$

and f is convex.

**Exercise 7.1.16.** Let X be a vector space,  $A \subset X$  convex and  $(f_{\lambda})_{{\lambda} \in {\Lambda}} \subset \mathbb{R}^{A}$ . Suppose that for each  ${\lambda} \in {\Lambda}$ ,  $f_{\lambda}$  is convex. Then  $\sup_{{\lambda} \in {\Lambda}} f_{\lambda}$  is convex.

*Proof.* Define  $f = \sup_{\lambda \in \Lambda} f_{\lambda}$ . Let  $x, y \in A, t \in [0, 1]$  and  $\lambda \in \Lambda$ . Then

$$f_{\lambda}(tx + (1-t)y) \le tf_{\lambda}(x) + (1-t)f_{\lambda}(y)$$
  
 
$$\le tf(x) + (1-t)f(y)$$

Since  $\lambda \in \Lambda$  is arbitrary,  $f(tx + (1-t)y) \leq tf(x) + (1-t)f(y)$ .

**Exercise 7.1.17.** Let X be a normed vector space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex and  $x_0 \in A$ . If f is continuous at  $x_0$ , then f is locally Lipschitz at  $x_0$ .

**Hint:** Given  $x_1, x_2$  near  $x_0$  Choose a z near  $x_0$  s.t.  $x_1$  is a convex combination of  $x_2$  and z. Then repeat but with  $x_2$  as a convex combination of  $x_1$  and z

*Proof.* By continuity, f is locally bounded at  $x_0$ . So there exist  $M, \delta > 0$  such that  $B(x_0, \delta) \subset A$  and for each  $x \in B(x_0, \delta), |f(x)| \leq M$ . Put  $\delta' = \frac{\delta}{2}$  and choose  $U = B(x_0, \delta')$ . Then  $U \subset A$  and  $U \in \mathcal{N}_{x_0}$ .

Let  $x_1, x_2 \in U$ . Suppose that  $x_1 \neq x_2$ . Define  $\alpha = ||x_1 - x_2|| > 0$ ,  $p = \frac{\alpha}{\alpha + \delta'}$ , q = 1 - p and  $z = p^{-1}(x_1 - qx_2)$ . Then  $x_1 = pz + qx_2$  and

$$||z - x_1|| = ||(p^{-1} - 1)x_1 - p^{-1}qx_2||$$

$$= \frac{1 - p}{p}\alpha$$

$$= \frac{\delta'}{\alpha}\alpha$$

$$= \delta'$$

Therefore

$$||z - x_0|| \le ||z - x_1|| + ||x_1 - x_0||$$
  
 $< \delta' + \delta'$   
 $= \delta$ 

So  $z \in B(x_0, \delta)$ , which implies that

$$f(z) - f(x_2) \le |f(z) - f(x_2)|$$
  
 $\le |f(z)| + |f(x_2)|$   
 $\le 2M$ 

Since  $x_1 = pz + qx_2$ , convexity of f implies that  $f(x_1) \leq pf(z) + qf(x_2)$ . Hence

$$f(x_1) - f(x_2) \le pf(z) - pf(x_2)$$

$$= p(f(z) - f(x_2))$$

$$\le p2M$$

$$= \frac{\alpha}{\alpha + \delta'} 2M$$

$$\le \alpha 2M$$

$$= 2M ||x_1 - x_2||$$

Similarly, choosing  $z = p^{-1}(x_2 - qx_1)$ , yields  $f(x_2) - f(x_1) \le 2M||x_1 - x_2||$  which implies that

$$|f(x_1) - f(x_2)| \le 2M||x_1 - x_2||$$

and f is Lipschitz on U.

#### 7.2. The Subdifferential.

**Exercise 7.2.1.** Let X be a Banach space,  $A \subset X$  open and convex,  $f: A \to \mathbb{R}$  convex,  $x_0 \in A$  and  $x \in X$ . Define  $T = \{t \in \mathbb{R} : x_0 + tx \in A\}$ . Then there exist  $a, b \in (0, \infty]$  such that T = (-a, b).

*Proof.* Continuity of scalar multiplication and addition implies that T is an open neighborhood of 0. Let t > 0 and  $s \in [0, t]$ . Then  $\frac{s}{t} \in [0, 1]$  and by convexity of A,  $x_0 + tx \in A$  implies that

$$x_0 + sx = \frac{s}{t}(x_0 + tx) + \left(1 - \frac{s}{t}\right)x_0$$

$$\in A$$

Thus  $[0,t] \subset T$ . Similarly,  $x_0 - tx \in A$  implies that  $[-t,0] \subset T$ . Define  $a,b \in (0,\infty]$  by  $a = \sup\{t > 0 : x_0 - tx \in A\}$  and  $b = \sup\{t > 0 : x_0 + tx \in A\}$ . Then (-a,b) = T.

**Definition 7.2.2.** Let X be a Banach space,  $A \subset X$  open and convex,  $f: A \to \mathbb{R}$  convex,  $x_0 \in A$  and  $x \in X$ . Define T as in the previous exercise and choose  $t_0 > 0$  such that  $(-t_0, t_0) \subset T$ . For  $t \in (0, t_0)$ , define the difference quotient  $q: (-t_0, t_0) \setminus \{0\} \to \mathbb{R}$  by

$$q(t) = \frac{f(x_0 + tx) - f(x_0)}{t}$$

**Exercise 7.2.3.** Let X be a Banach space,  $A \subset X$  open and convex,  $f: A \to \mathbb{R}$  convex,  $x_0 \in A$  and  $x \in X$ . Define  $t_0$  as above. Then

- (1) q(t) is increasing on  $(0, t_0)$
- (2) q(-t) decreasing on  $(0, t_0)$

**Hint:** As an example, look at the graph of  $f(x) = x^2$ . For the algebra, start at the desired end inequality and work backwards

Proof.

(1) Let  $s, t \in (0, t_0)$  and suppose that  $s \le t$ . Then  $x_0 + sx$ ,  $x_0 + tx \in A$ . Note that since  $0 < s \le t$ ,  $\frac{s}{t} \in (0, 1]$  and  $1 - \frac{s}{t} = \frac{t-s}{t} \in (0, 1]$ . Also, since A is convex, we have that

$$\left(\frac{t-s}{t}\right)x_0 + \left(\frac{s}{t}\right)(x_0 + tx) \in A$$

Convexity of f implies that

$$f(x_0 + sx) = f\left(\left(\frac{t-s}{t}\right)x_0 + \left(\frac{s}{t}\right)(x_0 + tx)\right)$$
  
$$\leq \left(\frac{t-s}{t}\right)f(x_0) + \left(\frac{s}{t}\right)f(x_0 + tx)$$

This implies that

$$tf(x_0 + sx) \le (t - s)f(x_0) + sf(x_0 + tx)$$

and after rearranging, we get

$$tf(x_0 + sx) - tf(x_0) \le sf(x_0 + tx) - sf(x_0)$$

and so finally, dividing both sides by st, we obtain

$$q(s) = \frac{f(x_0 + sx) - f(x_0)}{s}$$

$$\leq \frac{f(x_0 + tx) - f(x_0)}{t}$$

$$= q(t)$$

as desired.

(2) Similar to (1).

**Exercise 7.2.4.** Let X be a Banach space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex,  $x_0 \in A$  and  $x \in X$ . Define  $t_0$  as before. Then for each  $t \in (0, t_0)$ ,

$$q(-t) \le q(t)$$

**Hint:** for sufficiently small t, convexity of f implies that  $f(x_0) \leq \frac{1}{2}f(x_0 - 2tx) + \frac{1}{2}f(x_0 + 2tx)$ 

*Proof.* Choose  $t_0$  as in the previous exercise. Since convexity of f implies that for each  $t \in (0, t_0/2)$ ,

$$f(x_0) \le \frac{1}{2}f(x_0 - 2tx) + \frac{1}{2}f(x_0 + 2tx)$$

we have that for each  $t \in (0, t_0/2)$ ,

$$q(-2t) = \frac{f(x_0 - 2tx) - f(x_0)}{-2t}$$

$$\leq \frac{f(x_0 + 2tx) - f(x_0)}{2t}$$

$$= q(2t)$$

So for each  $t \in (0, t_0), q(-t) \leq q(t)$ .

**Exercise 7.2.5.** Let X be a Banach space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex and  $x_0 \in A$ . Then

- (1) f is left-hand and right-hand Gateaux differentiable at  $x_0$  with  $d^-f(x_0) \leq d^+f(x_0)$
- (2) for each  $x \in X$ ,  $d^-f(x_0)(x) = -d^+f(x_0)(-x)$

Proof.

(1) Let  $x \in X$ . Choose  $t_0 > 0$  as in the previous two exercises. Let  $t, u \in (0, t_0)$ . Choose  $s \in (0, \min(u, t))$ . The previous two exercises imply that

$$q(-u) \le q(-s)$$

$$\le q(s)$$

$$\le q(t)$$

and therefore q(t) is an upper bound for  $\{q(-u): u \in (0,t_0)\}$  and  $d^-f(x_0)(x) = \sup_{u \in (0,t_0)} q(-u)$  exists with  $d^-f(x_0)(x) \leq q(t)$ .

Since  $t \in (0, t_0)$  is arbitrary,  $d^-f(x_0)(x)$  is a lower bound for  $\{q(t) : t \in (0, t_0)\}$ . Therefore

$$d^+ f(x_0)(x) = \inf_{t \in (0, t_0)} q(t)$$

exists with  $d^+f(x_0)(x) \ge d^-f(x_0)(x)$ .

(2) By definition, we have

$$d^{-}f(x_{0})(x) = \lim_{t \to 0^{+}} \frac{f(x_{0} + -tx) - f(x_{0})}{-t}$$

$$= -\lim_{t \to 0^{+}} \frac{f(x_{0} + -tx) - f(x_{0})}{t}$$

$$= -d^{+}f(x_{0})(-x)$$

**Exercise 7.2.6.** Let X be a Banach space,  $A \subset X$  open and convex,  $f: A \to \mathbb{R}$  convex and  $x_0 \in A$ . Then  $d^+f(x_0): X \to \mathbb{R}$  is a sublinear functional.

*Proof.* Let  $x, y \in X$  and  $k \ge 0$ . If k = 0, then clearly

$$d^+f(x_0)(kx) = kd^+(x_0)(x)$$

If k > 0. Then

$$d^{+}f(x_{0})(kx) = \lim_{t \to 0^{+}} \frac{f(x_{0} + tkx) - f(x_{0})}{t}$$
$$= k \lim_{t \to 0^{+}} \frac{f(x_{0} + tkx) - f(x_{0})}{tk}$$
$$= kd^{+}f(x_{0})(x)$$

Define  $t_0 > 0$  as before and let  $t \in (0, \frac{t_0}{2})$ . Note that

$$x_0 + tx + ty = \frac{1}{2}(x_0 + 2tx) + \frac{1}{2}(x_0 + 2ty)$$

Convexity of f implies that

$$f(x_0 + tx + ty) \le \frac{1}{2}f(x_0 + 2tx) + \frac{1}{2}f(x_0 + 2ty)$$

which implies that

$$\frac{f(x_0 + tx + ty) - f(x_0)}{t} \le \frac{f(x_0 + 2tx) - f(x_0)}{2t} + \frac{f(x_0 + 2ty) - f(x_0)}{2t}$$

Therefore

$$d^{+}f(x_{0})(x+y) = \lim_{t \to 0^{+}} \frac{f(x_{0} + t(x+y)) - f(x_{0})}{t}$$

$$= \lim_{t \to 0^{+}} \frac{f(x_{0} + tx + ty) - f(x_{0})}{t}$$

$$\leq \lim_{t \to 0^{+}} \left[ \frac{f(x_{0} + 2tx) - f(x_{0})}{2t} + \frac{f(x_{0} + 2ty) - f(x_{0})}{2t} \right]$$

$$= \lim_{t \to 0^{+}} \frac{f(x_{0} + 2tx) - f(x_{0})}{2t} + \lim_{t \to 0^{+}} \frac{f(x_{0} + 2ty) - f(x_{0})}{2t}$$

$$= d^{+}f(x_{0})(x) + d^{+}f(x_{0})(y)$$

**Exercise 7.2.7.** Let X be a Banach space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex and  $x_0 \in A$ . Then for each  $x \in A$ ,

$$d^+f(x_0)(x - x_0) \le f(x) - f(x_0)$$

*Proof.* Let  $x \in A$ . Define  $T = \{t \in \mathbb{R} : x_0 + t(x - x_0) \in A\}$  similarly to earlier. Clearly  $1 \in T$  and

$$d^{+}f(x_{0})(x - x_{0}) = \inf_{t \in (0,1]} \frac{f(x_{0} + t(x - x_{0})) - f(x_{0})}{t}$$
  
$$\leq f(x) - f(x_{0})$$

**Exercise 7.2.8.** Let X be a Banach space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex and  $x_0 \in A$ . If f is continuous at  $x_0$ , then  $d^+f(x_0)$  is Lipschitz (equivalently bounded).

*Proof.* Suppose that f is continuous at  $x_0$ . A previous exercise about convex functions tells us that f is locally Lipschitz at  $x_0$ , so there exists  $\delta, M > 0$  such that for each  $x_1, x_2 \in B(x_0, \delta)$ ,  $|f(x_1) - f(x_2)| \le M||x_1 - x_2||$ . Let  $x \in X$  and define  $t_0 = \frac{\delta}{||x||+1}$  so that for each  $t \in (0, t_0)$ ,

$$||(x_0 + tx) - x_0|| = t||x||$$

$$\leq t_0||x||$$

$$= \frac{\delta||x||}{||x|| + 1}$$

$$< \delta$$

and  $x_0 + tx \in B(x_0, \delta)$ . Then for each  $t \in (0, t_0)$ ,

$$d^{+}f(x_{0})(x) \leq \frac{f(x_{0} + tx) - f(x_{0})}{t}$$

$$\leq \frac{|f(x_{0} + tx) - f(x_{0})|}{t}$$

$$\leq t^{-1}M||(x_{0} + tx) - x_{0}||$$

$$= M||x||$$

Thus  $d^+f(x_0)$  is a bounded sublinear functional and a previous exercise in the section on sublinear functionals implies this is equivalent to  $d^+f(x_0)$  being Lipschitz.

**Exercise 7.2.9.** Let X be a Banach space,  $A \subset X$  open and convex,  $f: A \to \mathbb{R}$  convex and  $x_0 \in A$ . If f is continuous at  $x_0$ , then there exists  $\phi \in X^*$  such that  $\phi \leq d^+f(x_0)$ .

Proof. Suppose that f is continuous at  $x_0$ . The previous exercise implies that  $d^+f(x_0)$  is Lipschitz (equivalently bounded). A previous exercise in the section discussing sublinear functionals tells us that boundedness of  $d^+f(x_0)$  implies that there exists  $\phi \in X^*$  such that  $\phi \leq d^+f(x_0)$ .

## Definition 7.2.10. Subdifferential:

Let X be a Banach space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex and  $x_0 \in A$ . We define the **subdifferential of** f **at**  $x_0$ , denoted  $\partial f(x_0)$ , to be

$$\partial f(x_0) = \{ \phi \in X^* : \text{for each } x \in A, f(x_0) + \phi(x - x_0) \le f(x) \}$$

**Exercise 7.2.11.** Let X be a Banach space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex and  $x_0 \in A$ . If f is continuous at  $x_0$ , then  $\partial f(x_0) \neq \emptyset$ .

*Proof.* Suppose that f is continuous at  $x_0$ . The previous exercise tells us that there exists  $\phi \in X^*$  such that  $\phi \leq d^+ f(x_0)$ . Let  $x \in A$ . A previous exercise implies that

$$\phi(x - x_0) \le d^+ f(x_0)(x - x_0) \le f(x) - f(x_0)$$

Then  $f(x_0) + \phi(x - x_0) \le f(x)$ .

**Exercise 7.2.12.** Let X be a Banach space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex,  $\phi \in X^*$  and  $x_0 \in A$ . Then

(1) for each  $x \in A$ ,

$$\phi(x - x_0) \le f(x) - f(x_0)$$

iff

$$\phi \le d^+ f(x_0)$$

(2) 
$$\partial f(x_0) = \{ \phi \in X^* : \phi \le d^+ f(x_0) \}$$

Proof.

(1) Suppose that for each  $x \in A$ ,  $\phi(x - x_0) \le f(x) - f(x_0)$ . Let  $x \in X$ . Define  $t_0$  as before. Then for each  $t \in (0, t_0)$ ,

$$t\phi(x) = \phi((x_0 + tx) - x_0)$$
  

$$\leq f(x_0 + tx) - f(x_0)$$

This implies that  $\phi(x) \leq d^+ f(x_0)(x)$ .

Conversely, suppose that  $\phi \leq d^+ f(x_0)$ . Let  $x \in A$ . A previous exercise implies that,

$$\phi(x - x_0) \le d^+ f(x_0)(x - x_0) \le f(x) - f(x_0)$$

(2) Clear.

**Exercise 7.2.13.** Let X be a Banach space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex and  $x_0 \in A$ . If f is continuous at  $x_0$ , then the following are equivalent:

(1) f is Gateaux differentiable at  $x_0$ 

- (2)  $d^+f(x_0)$  is linear
- (3)  $\#\partial f(x_0) = 1$

*Proof.* Suppose that f is continuous at  $x_0$ . Then  $d^+f(x_0)$  is Lipschitz and bounded.

 $\bullet$  (1)  $\Longrightarrow$  (2):

Suppose that f is Gateaux differentiable at  $x_0$ . Let  $x \in X$ . Then a previous exercise implies that

$$-df^{+}(x_{0})(-x) = df^{-}f(x_{0})(x)$$
$$= df^{+}f(x_{0})(x)$$

An exercise in the section on sublinear functionals implies that  $df^+f(x_0)$  is linear.

- (2)  $\Longrightarrow$  (3): Suppose that  $df^+f(x_0)$  is linear. Let  $\phi \in \partial f(x_0)$ . The previous exercise implies that  $\phi \leq df^+f(x_0)$ . Equivalence of linearity in the section on sublinear functionals implies that  $d^+f(x_0) = \phi$ .
- (3)  $\Longrightarrow$  (1): Suppose that  $\#\partial f(x_0) = 1$ . Since  $\partial f(x_0) = \{\phi \in X^* : \phi \leq d^+f(x_0)\}$ , equivalence of linearity in the section on sublinear functionals implies that  $d^+f(x_0)$  is linear. This implies that  $d^+f(x_0) = d^-f(x_0)$  and which implies that f is Gateaux differentiable at  $x_0$ .

**Exercise 7.2.14.** Let X be a Banach space,  $A \subset X$  open and convex,  $f, g : A \to \mathbb{R}$  convex,  $\lambda \geq 0$  and  $x_0 \in A$ . Then

$$\partial f(x_0) + \lambda \partial g(x_0) \subset \partial [f + \lambda g](x_0)$$

Proof. Let  $\zeta \in \partial f(x_0) + \lambda \partial g(x_0)$ . Then there exist  $\phi \in \partial f(x_0)$  and  $\psi \in \partial g(x_0)$  such that  $\zeta = \phi + \lambda \psi$ . A previous exercise implies that  $\phi \leq d^+ f(x_0)$  and  $\lambda \psi \leq \lambda d^+ g(x_0) = d^+ [\lambda g](x_0)$ . Hence

$$\zeta = \phi + \lambda \psi$$
  

$$\leq d^+ f(x_0) + d^+ [\lambda g](x_0)$$
  

$$= d^+ [f + \lambda g](x_0)$$

So  $\zeta \in \partial [f + \lambda g](x_0)$ 

**Exercise 7.2.15.** Let X be a Banach space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex and  $x_0 \in A$ . If f is continuous at  $x_0$ , then f has a global minimum point at  $x_0$  iff  $0 \in \partial f(x_0)$ .

*Proof.* Suppose that f has a global minimum point at  $x_0$ . Let  $x \in X$ . Then

$$d^{+}f(x_{0})(x) = \lim_{t \to 0^{+}} \frac{f(x_{0} + tx) - f(x_{0})}{t}$$
  
 
$$\geq 0$$

So  $0 \le df^+(x_0)$  and  $0 \in \partial f(x_0)$ .

Conversely, suppose that  $0 \in \partial f(x_0)$ . Let  $x \in A$ . Then

$$0 = 0(x - x_0)$$
  

$$\leq f(x) - f(x_0)$$

So that  $f(x_0) \leq f(x)$  which implies that f has a global minimum point at  $x_0$ .

**Exercise 7.2.16.** et X be a Banach space,  $A \subset X$  open and convex,  $f : A \to \mathbb{R}$  convex and  $x_0 \in A$ . If f is Frechet differentiable at  $x_0$ , then  $\partial f(x_0) = \{Df(x_0)\}.$ 

Proof. Clear.  $\Box$ 

**Exercise 7.2.17.** Let X be a Banach space,  $A \subset X$  open and convex,  $f: A \to \mathbb{R}$  convex and  $x_0 \in A$ . Suppose that f is Frechet differentiable at  $x_0$ . If  $Df(x_0) = 0$ , then f has a global minimum point at  $x_0$ .

*Proof.* Suppose that  $Df(x_0) = 0$ . Since  $\partial f(x_0) = \{Df(x_0)\}$ , a previous exercise implies that f has a global minimum point at  $x_0$ .

**Exercise 7.2.18.** Let X be a Banach space,  $A \subset X$  open and convex,  $f: A \to \mathbb{R}$  convex and  $x_0 \in A$ . Suppose that f is Frechet differentiable at  $x_0$ . Then for each  $x \in A$ ,  $f(x) \ge f(x_0) + Df(x_0)(x - x_0)$ 

Proof. Since  $Df(x_0) \in \partial f(x_0)$ , for each  $x \in A$ ,  $Df(x_0)(x - x_0) \le f(x) - f(x_0)$ .

**Exercise 7.2.19.** Let X be a Banach space,  $A \subset X$  open and convex, and  $f \in C^2(A)$ . Then f is convex iff for each  $x_0 \in A$ ,  $D^2 f(x_0)$  is positive semidefinite.

**Hint:** Define  $g: A \to \mathbb{R}$  by  $g(x) = f(x) - Df(x_0)(x - x_0)$  and show g is convex and use Taylor's Theorem

Proof. Suppose that f is convex. Let  $x_0 \in X$ . Define  $g: A \to \mathbb{R}$  by  $g(x) = f(x) - Df(x_0)(x-x_0)$ . Since g is the sum of a convex function and an affine function, g is convex. Since  $f \in C^2(A)$ , we have that  $g \in C^2(A)$  and it is straightforward to show that for each  $x \in A$ ,  $Dg(x) = Df(x) - Df(x_0)$  and  $D^2g(x) = D^2f(x)$ . In particular,  $Dg(x_0) = 0$ . Hence g has a global minimum point at  $x_0$ . This implies that  $D^2f(x_0)$  is positive semidefinite. Conversely, suppose that for each  $x_0 \in A$ ,  $D^2f(x_0)$  is positive semidefinite. Let

# 7.3. Conjugacy.

**Definition 7.3.1.** Let X be a Banach space,  $A \subset X$  and  $f : A \to \mathbb{R}$ . Define  $A^* \subset X^*$  and  $f^* : A^* \to \mathbb{R}$  by

$$A^* = \left\{ \phi \in X^* : \sup_{x \in A} \left[ \phi(x) - f(x) \right] < \infty \right\}$$

and

$$f^*(\phi) = \sup_{x \in A} \left[ \phi(x) - f(x) \right]$$

If X is a Hilbert space, we may define  $A^* \subset X$  and  $f^* : A^* \to \mathbb{R}$  via the Riesz representation theorem by

$$A^* = \left\{ y \in X : \sup_{x \in A} \left[ \langle y, x \rangle - f(x) \right] < \infty \right\}$$

and  $f^*: A^* \to \mathbb{R}$  and

$$f^*(y) = \sup_{x \in A} \left[ \langle y, x \rangle - f(x) \right]$$

**Exercise 7.3.2.** Let X be a Banach space,  $A \subset X$  and  $f: A \to \mathbb{R}$ . Then  $f^*$  is convex.

*Proof.* For  $x \in A$ , define  $g_x : X^* \to [\infty, \infty)$  by  $g_x(\phi) = \phi(x) - f(x)$ . Then for each  $x \in A$ ,  $g_x$  is convex since it is affine. Thus  $f^* = \sup_{x \in A} g_x$  is convex.

**Exercise 7.3.3.** Let X be a Banach space,  $A \subset X$  and  $f : A \to \mathbb{R}$ . Then for each  $x \in X$  and  $\phi \in X^*$ ,  $f(x) \ge \phi(x) - f^*(\phi)$ .

Proof. Clear 
$$\Box$$

Exercise 7.3.4.

Definition 7.3.5. Let

Definition 7.3.6.  $\partial f$ 

Exercise 7.3.7.

# 7.4. Functional Optimization.

**Exercise 7.4.1.** Let X be a Banach space,  $(S, \mathcal{S}, \mu)$  a measure space,  $A \subset X$ ,  $K \in L^0(A, \mathbb{R})$  and  $\Lambda \subset L^0(S, A) \cap \{f : S \to A : K \circ f \in L^1(\mu)\}$ . Suppose that A and  $\Lambda$  are convex. Define  $\phi : \Lambda \to \mathbb{R}$  by

$$\phi f = \int K \circ f d\mu$$

Then K is convex implies that  $\phi$  is convex.

*Proof.* Suppose that K is convex. Let  $t \in [0,1]$  and  $f,g \in \Lambda$ . Convexity of K implies that for each  $s \in S$ ,

$$K[tf(s) + (1-t)g(s)] \le tK[f(s)] + (1-t)K[g(s)]$$

So

$$K\circ [tf+(1-t)g]\leq tK\circ f+(1-t)K\circ g$$

Therefore

$$\begin{split} \phi[tf+(1-t)g] &= \int K \circ [tf+(1-t)g] d\mu \\ &\leq \int tK \circ f + (1-t)K \circ g d\mu \\ &= t \int K \circ f d\mu + (1-t) \int K \circ g d\mu \\ &= t \phi f + (1-t) \phi g \end{split}$$

and  $\phi$  is convex.

## 8. Topological Groups

### 8.1. Introduction.

**Definition 8.1.1.** Let G be a group and  $\tau$  a topology on G. Then  $(G, \tau)$  is said to be a **topological group** if the maps  $(g, h) \mapsto gh$  and  $g \mapsto g^{-1}$  are continuous.

# 8.2. Automorphism Groups of Metric Spaces.

**Definition 8.2.1.** Let  $(X, \tau)$  be a topological space. Define

$$Aut(X) = \{ \sigma : X \to X : \sigma \text{ is a homeomorphism} \}$$

**Exercise 8.2.2.** Let (X, d) be a compact metric space. Then  $(Aut(X), d_u)$  is a topological group.

*Proof.* Let  $(\sigma_n)_{n\in\mathbb{N}}$ ,  $(\tau_n)_{n\in\mathbb{N}}\subset \operatorname{Aut}(X)$  and  $\sigma,\tau\in\operatorname{Aut}(X)$ . Suppose that  $\sigma_n\stackrel{\mathrm{u}}{\to}\sigma$  and  $\tau_n\stackrel{\mathrm{u}}{\to}\tau$ .

(1) Let  $\epsilon > 0$ . Since X is compact and  $\sigma$  is continuous,  $\sigma$  is uniformly continuous. Then there exists  $\delta > 0$  such that for each  $x, y \in X$ ,  $d(x, y) < \delta$  implies that  $d(\sigma(x), \sigma(y)) \le \epsilon/2$ . Choose  $N_{\sigma} \in \mathbb{N}$  such that for each  $n \in \mathbb{N}$ ,  $n \ge \mathbb{N}$  implies that  $d_u(\sigma_n, \sigma) < \epsilon/2$ . Choose  $N_{\tau} \in \mathbb{N}$  such that for each  $n \in \mathbb{N}$ ,  $n \ge \mathbb{N}$  implies that  $d_u(\tau_n, \tau) < \delta$ . Put  $N = \max(N_{\sigma}, N_{\tau})$ . Let  $n \in \mathbb{N}$  and  $x \in X$ . Suppose that  $n \ge N$ . Then

$$d(\sigma_n \circ \tau_n(x), \sigma \circ \tau(x)) \le d(\sigma_n(\tau_n(x)), \sigma(\tau_n(x))) + d(\sigma(\tau_n(x)), \sigma(\tau(x)))$$

$$< \epsilon/2 + \epsilon/2$$

$$= \epsilon$$

So  $d_u(\sigma_n \circ \tau_n, \sigma \circ \tau) \leq \epsilon$  and  $\circ : \operatorname{Aut}(X)^2 \to \operatorname{Aut}(X)$  is continuous.

(2) Suppose that  $\sigma = \mathrm{id}_X$ . Let  $\epsilon > 0$ . Then there exists  $N \in \mathbb{N}$  such that for each  $n \in \mathbb{N}$ ,  $n \geq N$  implies that  $d_u(\sigma_n, \mathrm{id}_X) < \epsilon$ . Let  $n \in \mathbb{N}$ . Suppose that  $n \geq N$ . Then

$$\sup_{x \in X} d(\sigma_n^{-1}(x), x) = \sup_{x \in \sigma_n(X)} d(\sigma_n^{-1}(x), x)$$

$$= \sup_{x \in X} d(\sigma_n^{-1}(\sigma_n(x)), \sigma_n(x))$$

$$= \sup_{x \in X} d(x, \sigma_n(x))$$

$$< \epsilon$$

So  $\sigma_n^{-1} \xrightarrow{\mathrm{u}} \mathrm{id}_X$ . Now suppose that  $\sigma \neq \mathrm{id}_X$ . Since  $\sigma_n \xrightarrow{\mathrm{u}} \sigma$ , part (1) implies that  $\sigma^{-1} \circ \sigma_n \xrightarrow{\mathrm{u}} \mathrm{id}_X$ . Applying the result from above, we get that  $\sigma_n^{-1} \circ \sigma \xrightarrow{\mathrm{u}} \mathrm{id}_X$ . Applying part (1) again implies that  $\sigma_n^{-1} \xrightarrow{\mathrm{u}} \sigma^{-1}$ . So the map  $\sigma \mapsto \sigma^{-1}$  is continuous.

Hence Aut(X) is a topological group.

**Definition 8.2.3.** Let (X, d) be a metric space. Define

$$\operatorname{Aut}(X,d) = \{ \sigma : X \to X : \sigma \text{ is an isometric isomorphism} \}$$

**Exercise 8.2.4.** Let (X, d) be a compact metric space. Then  $(\operatorname{Aut}(X, d), d_u)$  is a compact subgroup of  $(\operatorname{Aut}(X), d_u)$ .

*Proof.* Clearly,  $(Aut(X, d), d_u)$  is a topological subgroup. To show compactness, use the Arzela Ascoli theorem.

**Definition 8.2.5.** Let  $(X, \tau)$  be a topological space and  $\mu : \mathcal{B}(X) \to \mathbb{R}$  a Borel measure. Define

$$\operatorname{Aut}(X,\mu) = \{ \sigma \in \operatorname{Aut}(X) : \sigma_*\mu = \mu \}$$

**Exercise 8.2.6.** Let (X, d) be a compact metric space and  $\mu : \mathcal{B}(X) \to \mathbb{R}$  an outer-regular Borel measure. Then  $\operatorname{Aut}(X, \mu)$  is a closed subgroup of  $\operatorname{Aut}(X)$ .

*Proof.* It is clear that  $\operatorname{Aut}(X,\mu)$  is a subgroup of  $\operatorname{Aut}(X)$ . Let  $(\sigma_n)_{n\in\mathbb{N}}\subset\operatorname{Aut}(X,\mathcal{B}(X),\mu)$  and  $\sigma\in\operatorname{Aut}(X)$ . Suppose that  $\sigma_n\stackrel{\mathrm{u}}{\to}\sigma$ . Let  $E\subset X$  be closed,  $U\subset X$  open and suppose that  $E\subset U$ . An exercise in the section on metric spaces tells us that there exists  $N\in\mathbb{N}$  such that for each  $n\in\mathbb{N}, n\geq N$  implies that  $\sigma(E)\subset\sigma_n(U)$ . Then

$$\mu(\sigma(E)) \le \mu(\sigma_N(U))$$
$$= \mu(U)$$

Therefore, since  $\mu$  is outer regular,  $\mu(\sigma(E)) \leq \mu(E)$ . Since  $\sigma_n^{-1} \xrightarrow{\mathbf{u}} \sigma^{-1}$ , we may apply the above argument to obtain that

$$\mu(E) = \mu(\sigma^{-1}(\sigma(E)))$$

$$\leq \mu(\sigma(E))$$

Hence  $\mu(E) = \mu(\sigma(E))$ . Applying the whole argument above thus far to  $\sigma^{-1}$ , we see that  $\mu(E) = \mu(\sigma^{-1}(E))$ . Since  $E \subset X$  is an arbitrary closed set and  $\mathcal{B}(X) = \sigma(E \subset X : E \text{ is closed})$ , we have that  $\mu = \sigma_*\mu$ . Thus  $\sigma \in \operatorname{Aut}(X,\mu)$  which implies that  $\operatorname{Aut}(X,\mu)$  is closed.

**Definition 8.2.7.** Let (X, d) be a compact metric space and  $\mu : \mathcal{B}(X) \to \mathbb{R}$  an outer-regular Borel measure. Define  $\operatorname{Aut}(X, d, \mu) = \operatorname{Aut}(X, d) \cap \operatorname{Aut}(X, \mu)$ .

**Exercise 8.2.8.** Let (X, d) be a compact metric space and  $\mu : \mathcal{B}(X) \to \mathbb{R}$  an outer-regular Borel measure. Then  $\operatorname{Aut}(X, d, \mu)$  is compact.

*Proof.* Since  $\operatorname{Aut}(X,d)$  is compact and  $\operatorname{Aut}(X,\mu)$  is closed,  $\operatorname{Aut}(X,d,\mu)$  is compact.

# 8.3. Group Actions on Metric Spaces.

**Note 8.3.1.** For a set X, a group G and a (left) group action  $\phi : G \times X \to X$ , we will write  $\phi(g, x)$  as  $g \cdot x$ . We denote the projection map by  $\pi : X \to X/G$ .

**Definition 8.3.2.** Let (X, d) be a metric space, G a group, and  $\phi : G \times X \to X$  a group action. We define  $d_* : X/G \times X/G \to [0, \infty)$  by

$$d_*(o_x, o_y) = \inf_{\substack{a \in o_x \\ b \in o_y}} d(a, b)$$

**Definition 8.3.3.** Let (X, d) be a metric space, G a group, and  $\phi : G \times X \to X$  a group action. Then  $\phi$  is said to be an **isometric group action** if for each  $g \in G$ , the map  $x \mapsto g \cdot x$  is an isometry.

**Exercise 8.3.4.** Let (X, d) be a metric space, G a group, and  $\phi : G \times X \to X$  an isometric group action. Then for each  $x, y \in X$ ,

$$d_*(o_x, o_y) = \inf_{g \in G} d(g \cdot x, y)$$

*Proof.* Let  $x, y \in X$ ,  $a \in o_x$  and  $b \in o_y$ . Then there exists there exists  $g_x, g_y \in G$  such that  $a = g_x \cdot x$  and  $b = g_y \cdot y$ . Set  $g = g_y^{-1}g_x$ . Since the map  $z \mapsto g_y^{-1} \cdot z$  is an isometry,

$$d(a,b) = d(g_x \cdot x, g_y \cdot y)$$
$$= d(g_y^{-1} g_x \cdot x, y)$$
$$= d(g \cdot x, y)$$

Let  $\epsilon > 0$ . Then there exist  $a^* \in o_x$  and  $b^* \in o_y$  such that  $d(a^*, b^*) < d_*(o_x, o_y) + \epsilon$ . The above argument implies that there exists  $g^* \in G$  such that

$$\inf_{g \in G} d(g \cdot x, y) \le d(g^* \cdot x, y)$$

$$= d(a^*, b^*)$$

$$< d_*(o_x, o_y) + \epsilon$$

Since  $\epsilon > 0$  is arbitrary,

$$\inf_{g \in G} d(g \cdot x, y) \le d_*(o_x, o_y)$$

Conversely, since  $\{(g \cdot x, y) : g \in G\} \subset \{(a, b) : a \in o_x, b \in o_y\}$ , we have that

$$\inf_{g \in G} d(g \cdot x, y) \ge d_*(o_x, o_y)$$

**Exercise 8.3.5.** Let (X, d) be a metric space, G a group, and  $\phi : G \times X \to X$  an isometric group action. Then for each  $x, y, z \in X$ ,

$$d_*(o_x, o_y) \le d_*(o_x, o_z) + d_*(o_z, o_y)$$

*Proof.* Let  $x, y, z \in X$ . An exercise in section (2.1) implies that  $d(o_x, o_y) \leq d(o_x, z) + d(z, o_y)$ . The previous exercise implies that

$$d(o_x, z) = \inf_{a \in o_x} d(a, z)$$
$$= \inf_{g \in G} d(g \cdot x, z)$$
$$= d_*(o_x, o_z)$$

Similarly,  $d(z, o_y) = d_*(o_z, o_y)$ . Then

$$d(o_x, o_y) \le d(o_x, z) + d(z, o_y)$$
  
=  $d_*(o_x, o_z) + d_*(o_z, o_y)$ 

**Exercise 8.3.6.** Let (X, d) be a metric space, G a group, and  $\phi : G \times X \to X$  an isometric group action. If for each  $x \in X$ ,  $o_x$  is closed, then for each  $x, y \in X$ ,  $d_*(o_x, o_y) = 0$  implies that  $o_x = o_y$ .

Proof. Suppose that for each  $x \in X$ ,  $o_x$  is closed. Let  $x, y \in X$ . Suppose that  $d_*(o_x, o_y) = 0$ . Then  $\inf_{g \in G} d(g \cdot x, y) = 0$ . Hence there exists  $(g_n)_{n \in \mathbb{N}} \subset G$  such that  $g_n \cdot x \to y$ . Since  $(g_n \cdot x)_{n \in \mathbb{N}} \subset o_x$  and  $o_x$  is closed,  $y \in o_x$ . Thus  $o_x = o_y$ .

**Exercise 8.3.7.** Let (X, d) be a metric space, G a group, and  $\phi : G \times X \to X$  an isometric group action. If for each  $x \in X$ ,  $o_x$  is closed, then  $d_*$  is a metric on X/G.

*Proof.* Clear by preceding exercises.

**Exercise 8.3.8.** Let (X, d) be a metric space,  $(G, \tau)$  a topological group, and  $\phi : G \times X \to X$  an isometric group action. Suppose that G is compact and for each  $x \in X$ , the map  $g \mapsto g \cdot x$  is continuous. Then  $d_*$  is a metric on X/G.

*Proof.* Let  $x \in X$ . Since G is compact and the map  $g \mapsto g \cdot x$  is continuous,  $o_x = G \cdot x$  is compact and therefore closed. The previous exercise implies that  $d_*$  is a metric.

**Exercise 8.3.9.** Let (X, d) be a metric space, G a group, and  $\phi : G \times X \to X$  an isometric group action. Suppose that  $d_*$  is a metric on X/G. Then the projection map  $\pi : X \to X/G$  is Lipschitz and therefore continuous.

*Proof.* Let  $x, y \in X$ . Then

$$d_*(\pi(x), \pi(y)) = d_*(o_x, o_y)$$
$$= \inf_{g \in G} d(g \cdot x, y)$$
$$\leq d(x, y)$$

**Exercise 8.3.10.** Let (X, d) be a metric space, G a group, and  $\phi : G \times X \to X$  an isometric group action. Suppose that  $d_*$  is a metric on X/G. Let  $(x_n)_{n \in \mathbb{N}} \subset X$  and  $x \in X$ . Then  $o_{x_n} \xrightarrow{d_*} o_x$  iff there exists a sequence  $(g_n)_{n \in \mathbb{N}}$  such that  $g_n \cdot x_n \xrightarrow{d} x$ .

*Proof.* Suppose that  $o_{x_n} \xrightarrow{d_*} o_x$ . For  $n \in \mathbb{N}$ , choose  $g_n \in G$  such that  $d(g_n \cdot x_n, x) < d(o_{x_n}, o_x) + 2^{-n}$ . Then  $d(g_n \cdot x_n, x) \to 0$  and  $g_n \cdot x_n \xrightarrow{d_*} x$ .

Conversely, suppose that that there exists a sequence  $(g_n)_{n\in\mathbb{N}}$  such that  $g_n\cdot x_n\stackrel{d}{\to} x$ . Since  $\pi:X\to X/G$  is continuous, we have that

$$g_n \cdot x_n \xrightarrow{d} x \implies \pi(g_n \cdot x_n) \xrightarrow{d_*} \pi(x)$$
  
 $\implies o_{x_n} \xrightarrow{d_*} o_x$ 

**Exercise 8.3.11.** Let X be a set,  $d_1, d_2 : X^2 \to [0, \infty)$  metrics, G a group and  $\phi : G \times X \to X$  an isometric group action. Suppose that  $d_1$  and  $d_2$  are topologically equivalent.

- (1) Then  $d_{1*}$  is a metric on X/G iff  $d_{2*}$  is a metric on X/G
- (2) If  $d_{1*}$  and  $d_{2*}$  are metrics, then  $d_{1*}$  and  $d_{2*}$  are topologically equivalent.

Proof.

- (1)  $\longrightarrow$  Suppose that  $d_{1*}$  is a metric. Let  $x, y \in X$ . Suppose that  $d_{2*}(o_x, o_y) = 0$ . Then there exist  $(g_n)_{n \in \mathbb{N}} \subset G$  such that  $d_2(g_n \cdot x, y) \to 0$ . Since  $d_1$  and  $d_2$  are topologically equivalent,  $d_1(g_n \cdot x, y) \to 0$ . Thus  $d_{1*}(o_x, o_y) = 0$ . Since  $d_{1*}$  is a metric,  $o_x = o_y$ . Hence  $d_{2*}$  is a metric.
  - $\bullet \iff \text{Similar}.$
- (2) Suppose that  $d_{1*}$  and  $d_{2*}$  are metrics. Let  $(o_{x_n})_{n\in\mathbb{N}}\subset X/G$  and  $o_x\in X/G$ .
  - Suppose that  $o_{x_n} \xrightarrow{d_{1*}} o_x$ . Then there exists a sequence  $(g_n)_{n \in \mathbb{N}}$  such that  $g_n \cdot x_n \xrightarrow{d_1} x$ . Since  $d_1$  and  $d_2$  are topologically equivalent,  $g_n \cdot x_n \xrightarrow{d_2} x$ . This implies that  $o_{x_n} \xrightarrow{d_{2*}} o_x$ .
  - Suppose that  $o_{x_n} \xrightarrow{d_{2*}} o_x$ . Then similarly to above,  $o_{x_n} \xrightarrow{d_{1*}} o_x$ .

**Exercise 8.3.12.** Let X be a set,  $d_1, d_2 : X^2 \to [0, \infty)$  metrics on X, G a group and  $\phi : G \times X \to X$  an isometric group action. Suppose that  $d_1$  and  $d_2$  are equivalent. If  $d_{1*}$  and  $d_{2*}$  are metrics, then  $d_{1*}$  and  $d_{2*}$  are equivalent.

*Proof.* Suppose that  $d_{1*}$  and  $d_{2*}$  are metrics. Since  $d_1 d_2$  are equivalent, there exist  $C_1, C_2 > 0$  such that for each  $x, y \in X$ ,  $C_1 d_1(x, y) \le d_2(x, y) \le C_2 d_1(x, y)$ . Let  $x, y \in X$ . Then

$$C_1 d_{1*}(o_x, o_y) = C_1 \inf_{g \in G} d_1(g \cdot x, y)$$

$$= \inf_{g \in G} C_1 d_1(g \cdot x, y)$$

$$\leq \inf_{g \in G} d_2(g \cdot x, y)$$

$$= d_{2*}(o_x, o_y)$$

and

$$d_{2*}(o_x, o_y) = \inf_{g \in G} d_2(g \cdot x, y)$$

$$\leq \inf_{g \in G} C_2 d_1(g \cdot x, y)$$

$$= C_2 \inf_{g \in G} d_1(g \cdot x, y)$$

$$= C_2 d_{1*}(o_x, o_y)$$

So that  $C_1d_{1*} \leq d_{2*} \leq C_2d_{1*}$ 

## 8.4. Fundamental Examples.

**Exercise 8.4.1.** Consider the metric space  $(\mathbb{C}^{n\times d}, \|\cdot\|_F)$ , topological group  $(U_d, \|\cdot\|_F)$  and the (right) action  $X\cdot U=XU$ . Then this action is continuous,  $U_d$  is compact and for each  $U\in U_d$ , the map  $X\mapsto XU$  is an isometry. Thus  $d_*$  is a metric on  $\mathbb{C}^{n\times d}/U_d$ .

Proof. Clear. 
$$\Box$$

**Exercise 8.4.2.** Let X be a compact metric space and  $\mu : \mathcal{B}(X) \to [0, \infty]$  a Borel measure. Define the (right) group action  $L^1(\mu) \times \operatorname{Aut}(X, \mu) \to L^1(\mu)$  by

$$f \cdot \sigma = f \circ \sigma$$

Then for each  $\sigma \in \operatorname{Aut}(X, \mu)$ , the map  $f \mapsto f \cdot \sigma$  is an isometry.

*Proof.* Let  $\sigma \in \operatorname{Aut}(X, \mu)$  and  $f \in L^1(\mu)$ . Then

$$||f \cdot \sigma||_1 = \int_X |f \circ \sigma| d\mu$$

$$= \int_X |f| \circ \sigma d\mu$$

$$= \int_{\sigma(X)} |f| d\sigma_* \mu$$

$$= \int_{\sigma(X)} |f| d\mu$$

$$= \int_X |f| d\mu$$

$$= ||f||_1$$

**Exercise 8.4.3.** Let (X,d) be a compact metric space and  $\mu: \mathcal{B}(X) \to \mathbb{R}$  an outer-regular Borel measure. Then

## 9. Appendix

# 9.1. Summation.

**Definition 9.1.1.** Let  $f: X \to [0, \infty)$ , Then we define

$$\sum_{x \in X} f(x) := \sup_{\substack{F \subset X \\ F \text{ finite}}} \sum_{x \in F} f(x)$$

This definition coincides with the usual notion of summation when X is countable. For  $f: X \to \mathbb{C}$ , we can write f = g + ih where  $g, h: X \to \mathbb{R}$ . If

$$\sum_{x \in X} |f(x)| < \infty,$$

then the same is true for  $g^+, g^-, h^+, h^-$ . In this case, we may define

$$\sum_{x \in X} f(x)$$

in the obvious way.

The following note justifies the notation  $\sum_{x \in X} f(x)$  where  $f: X \to \mathbb{C}$ .

Note 9.1.2. Let  $f: X \to \mathbb{C}$  and  $\alpha: X \to X$  a bijection. If  $\sum_{x \in X} |f(x)| < \infty$ , then  $\sum_{x \in X} f(\alpha(x)) = \sum_{x \in X} f(x)$ .

# 9.2. Asymptotic Notation.

**Definition 9.2.1.** Let X be a topological space, Y, Z be normed vector spaces,  $f: X \to Y$ ,  $g: X \to Z$  and  $x_0 \in X \cup \{\infty\}$ . Then we write

$$f = o(g)$$
 as  $x \to x_0$ 

if for each  $\epsilon > 0$ , there exists  $U \in \mathcal{N}_{x_0}$  such that for each  $x \in U$ ,

$$||f(x)|| \le \epsilon ||g(x)||$$

**Exercise 9.2.2.** Let X be a topological space, Y, Z be normed vector spaces,  $f: X \to Y$ ,  $g: X \to Z$  and  $x_0 \in X \cup \{\infty\}$ . If there exists  $U \in \mathcal{N}_{x_0}$  such that for each  $x \in U \setminus \{x_0\}$ , g(x) > 0, then

$$f = o(g) \text{ as } x \to x_0 \text{ iff } \lim_{x \to x_0} \frac{\|f(x)\|}{\|g(x)\|} = 0$$

**Exercise 9.2.3.** Let X and Y a be normed vector spaces,  $A \subset X$  open and  $f: A \to Y$ . Suppose that  $0 \in A$ . If  $f(h) = o(\|h\|)$  as  $h \to 0$ , then for each  $h \in X$ , f(th) = o(|t|) as  $t \to 0$ .

*Proof.* Suppose that  $f(h) = o(\|h\|)$  as  $h \to 0$ . Let  $h \in X$  and  $\epsilon > 0$ . Choose  $\delta' > 0$  such that for each  $h' \in B(0, \delta')$ ,  $h' \in A$  and

$$||f(h')|| \le \frac{\epsilon}{||h|| + 1} ||h'||$$

Choose  $\delta > 0$  such that for each  $t \in B(0, \delta)$ ,  $th \in B(0, \delta')$ . Let  $t \in B(0, \delta)$ . Then

$$||f(th)|| \le \frac{\epsilon}{||h|| + 1} |t| ||h||$$

$$< \epsilon |t|$$

So f(th) = o(|t|) as  $t \to 0$ .

**Definition 9.2.4.** Let X be a topological space, Y, Z be normed vector spaces,  $f: X \to Y$ ,  $g: X \to Z$  and  $x_0 \in X \cup \{\infty\}$ . Then we write

$$f = O(g)$$
 as  $x \to x_0$ 

if there exists  $U \in \mathcal{N}_{x_0}$  and  $M \geq 0$  such that for each  $x \in U$ ,

$$||f(x)|| \le M||g(x)||$$

# References

- Introduction to Algebra
   Introduction to Analysis
   Introduction to Fourier Analysis
   Introduction to Measure and Integration