# **Normed Vector Spaces**

## **Vector Spaces**

Throughout,  $\mathbb{F} = \mathbb{R}$  or  $\mathbb{C}$ . A **vector space** over  $\mathbb{F}$  is a nonempty set V equipped with two operations: vector addition and scalar multiplication.

$$V \times V \xrightarrow{+} V$$
  $(v, w) \mapsto v + w$  Vector Addition  $F \times V \to V$   $(\alpha, v) \mapsto \alpha v$  Scalar Multiplication

The vector space is an Abelian group, where  $u, v, w \in V$  and  $\alpha, \beta \in \mathbb{F}$ , we have:

(i) 
$$u + (v + w) = (u + v) + w$$

(ii) 
$$\exists 0_v \in V$$
 with  $\forall v \in V$ ,  $0_v + v = v + 0_v = v$ 

(iii) 
$$(\forall v \in V)(\exists w \in V)$$
 with  $v + w = 0_v$ 

(iv) 
$$\forall v, w \in V, v + w = w + v$$

(v) 
$$\alpha(v+w) = \alpha v + \alpha w$$
,  $(\alpha + \beta)v = \alpha v + \beta v$ 

(vi) 
$$\alpha(\beta w) = (\alpha \beta) w$$

(vii) 
$$1 \cdot v = v$$

### Remarks:

- (a)  $0_V$  is unique and known as the zero vector.
- (b) The vector w in (iii) is unique, and denoted -v.

(c) 
$$0 \cdot v = 0_v$$

(d) 
$$(-1) \cdot v = -v$$

(e) Property (iv) follows from all the other axioms.

(f) For 
$$n \in \mathbb{N}$$
,  $n \cdot v = \underbrace{v + v + \dots + v}_{n \text{ times}}$ 

## **Subspaces**

Let V be a vector space over  $\mathbb{F}$ . A **subspace** is a nonempty subset  $W \subseteq V$  satisfying the following:

(i) 
$$w \in W, \alpha \in \mathbb{F} \to \alpha w \in W$$
.

(ii) 
$$w_1, w_2 \in W \Rightarrow w_1 + w_2 \in W$$
.

**Remark:**  $0_{\nu}$  is always a member of any subspace; a subspace is also a vector space.

## Proposition: Intersection of Subspaces

If  $\{W_i\}_{i\in I}$  is a family of subspaces of V, then,  $\bigcap W_i$  is a subspace of V.

### Proposition: Union of Subspaces

It is not the case that the union of subspaces of V also a subspace. For example, consider  $\mathbb{R}^2$  with the traditional vector space operations:

$$\begin{pmatrix} x \\ y \end{pmatrix} + \begin{pmatrix} x' \\ y' \end{pmatrix} = \begin{pmatrix} x + x' \\ y + y' \end{pmatrix}$$

$$\alpha \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} \alpha x \\ \alpha y \end{pmatrix}$$

If  $W_1, W_2 \in V$  are subspaces such that  $W_1 \cup W_2$  is a subspace, then  $W_1 \subseteq W_2$  or  $W_2 \subseteq W_1$ .

### Generated Subspaces

Let  $S \subseteq V$  be any subset of a vector space V. Then,

$$\operatorname{span}(S) = \left\{ \sum_{j=1}^n \alpha_j v_j \mid \alpha_1, \dots, \alpha_n \in \mathbb{F}, v_1, \dots, v_n \in S \right\}$$

### Remarks:

- $\operatorname{span}(S) \subseteq V$  is a subspace.
- $\operatorname{span}(S) = \bigcap W$ , where  $S \subseteq W$  and  $W \subseteq V$  is a subspace. Thus,  $\operatorname{span}(S)$  is the "smallest" subspace containing S, or the subspace generated by S.

#### Proposition: Quotient Group on Vector Space

Let V be a vector space, and let  $W \subseteq V$  is a subspace. Define  $u \sim_W v \leftrightarrow u - v \in W$ .

- (1)  $\sim_W$  is an equivalence relation.
- (2) If  $[v]_W$  denotes the equivalence class of v, then  $[v]_W = v + W = \{v + w | w \in W\}$ .
- (3)  $V/W := \{[v]_W | v \in V\}$  is a vector space with  $[v_1]_W + [v_2]_W = [v_1 + v_2]_W$  and  $\alpha[v]_W = [\alpha v]_W$ .

#### Proof of (1):

- Reflexive:  $u \sim_W u$ , since  $u u = 0 \in W$ .
- Transitive: Suppose  $u \sim_W v$ , and  $v \sim_W z$ . Then,  $u v \in W$ , and  $v z \in W$ . So,  $(u v) + (v z) \in W$ , so  $u z \in W$ . Whence,  $u \sim_W z$ .
- Symmetric: If  $u \sim_W v$ , then  $u v \in W$ , so  $-1 \cdot (u v) \in W$ , so  $v u \in W$ . Whence,  $v \sim_W u$ .

### Proof of (2):

$$[v]_{W} = \{u \in V \mid u \sim_{W} v\}$$

$$= \{u \in V \mid u - v \in W\}$$

$$= \{u \in V \mid u = v + w \text{ some } w \in W\}$$

$$= \{v + w \mid w \in W\}$$

$$= v + W$$

Proof of (3): Prove that the operation is well-defined.

#### **Bases**

Let V be a vector space and  $S \subseteq V$  be a subset.

- (1) S is said to be spanning for V if span(S) = V.
- (2) S is linearly independent if, for  $\sum_{i=1}^{n} \alpha_{j} v_{j} = 0_{v}$  with  $\alpha_{1}, \ldots, \alpha_{n} \in \mathbb{F}$ ,  $v_{1}, \ldots, v_{n} \in S$ , then  $\alpha_{1} = \alpha_{2} = \cdots = \alpha_{n} = 0$ .
- (3) S is a basis for V if S is linearly independent and spanning for V.

### Proposition: Existence of Basis

Every vector space admits a basis. If  $B_0 \subseteq V$  is linearly independent,  $\exists B \subseteq V$  such that B is a basis and  $B \supseteq B_0$ .

**Background:** A relation on a set X is a subset  $R \subseteq X \times X$ . If R is reflexive  $(x \sim x)$ , transitive  $(x \sim y, y \sim z \rightarrow x \sim z)$ , and antisymmetric  $(x \sim y, y \sim x \rightarrow x = y)$ , then R is an ordering, and we write  $x \leq y$ .

If  $\leq$  is an ordering of X such that  $\forall x, y \in X$ ,  $x \leq y$  or  $y \leq x$ , then  $\leq$  is a total (or linear) ordering.

Let  $\leq$  be an ordering of X, let  $Y \subseteq X$ . An upper bound for Y is an element  $u \in X$  such that  $y \leq u \ \forall y \in Y$ . A maximal element in X is an element  $m \in X$  such that  $x \in X$ ,  $x \geq m \to x = m$ .

**Example:**  $\mathbb N$  under the division ordering defines  $a \le b \Leftrightarrow a|b$ . If we want to find the maximal elements of  $A = \{2, 6, 9, 12\}$ , we would see that they are 9 and 12 (since no element of A can be divided by 9 and 12). Meanwhile,  $\mathbb N$  itself has no maximal elements.

This leads us to ask: given an ordered set,  $(X, \leq)$ , does X admit maximal elements.

**Zorn's Lemma (or Axiom):** Let  $(X, \leq)$  be an ordered set. Suppose that every totally ordered subset,  $Y \subseteq X$  has an upper bound in X. Then, X admits at least one maximal element.

The proof of Zorn's Lemma relies on the Axiom of Choice (and Zorn's Lemma is equivalent to the Axiom of Choice).

**Proof:** Let  $X = \{D \mid B_0 \subseteq D \subseteq V\}$  with D linearly independent. Since  $B_0 \subseteq X$ ,  $X \neq \emptyset$ . Define  $D, E \in X$ ,  $D \subseteq E \Leftrightarrow D \subseteq E$ . We will show that X has a maximal element.

Consider any totally ordered subset,  $Y = \{D_i\}_{i \in I}$ . Consider  $D = \bigcup D_i$ . Clearly,  $B_0 \subseteq D \subseteq V$ . Suppose  $\sum \alpha_k v_k = 0_V$  with  $v_1, \ldots, v_n \in D$ . Therefore,  $\exists D_j$  with  $v_1, \ldots, v_n \in D_j$  because Y is totally ordered. However, by definition,  $D_j$  is a linearly independent set — therefore,  $\alpha_k = 0$ . Thus, D is linearly independent.

Since D is linearly independent, and  $B_0 \subseteq D$ , it must be the case that  $D \in X$ . D is also an upper bound for Y. So, by Zorn's Lemma, X has a maximal element, B.

So,  $B_0 \subseteq B \subseteq V$ , B is independent, and B is maximal in X. We claim that B is a basis for V. Suppose toward contradiction that  $\exists v \in V$  such that  $v \notin \text{span}(B)$ . Consider  $B' = B \cup \{v\}$ .

Then,  $B_0 \subseteq B'$ , and B' is linearly independent — if  $\sum \alpha_k v_k + \alpha v = 0$ , where  $v_1, \ldots, v_n \in B$ , then either:

- If  $\alpha = 0$ , then  $\alpha_k v_k = 0 \Rightarrow \alpha_k = 0$ .
- If  $\alpha \neq 0$ , then  $\sum \alpha_k v_k = -\alpha v$ , which means  $v \in \text{span}(B)$ .  $\perp$

Thus, we have a linearly independent set, B', with  $B \subseteq B'$ , and  $B_0 \subseteq B'$ . Therefore,  $B' \in X$ . However, this contradicts the maximality of B. Therefore, span(B) = V, and B is a basis for V.

### **Examples: Vector Spaces**

(1) n-Dimensional Vectors:

$$\mathbb{F}^{n} = \left\{ \begin{pmatrix} x_{1} \\ \vdots \\ x_{n} \end{pmatrix} \mid x_{j} \in \mathbb{F} \right\}$$

$$\begin{pmatrix} x_{1} \\ \vdots \\ x_{n} \end{pmatrix} + \begin{pmatrix} y_{1} \\ \vdots \\ y+n \end{pmatrix} = \begin{pmatrix} x_{1}+y+1 \\ \vdots \\ x_{n}+y+n \end{pmatrix}$$

$$\alpha \begin{pmatrix} x_{1} \\ \vdots \\ x_{n} \end{pmatrix} = \begin{pmatrix} \alpha x_{1} \\ \vdots \\ \alpha x_{n} \end{pmatrix}$$

$$\beta = \{e_{1}, \dots, e_{n}\}$$

where  $e_i$  denotes the unit vector at position i.

(2)  $m \times n$  Matrices:

$$\mathbb{M}_{m,n}(\mathbb{F}) = \left\{ \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix} \mid a_{ij} \in \mathbb{F} \right\}$$
$$(a_{ij}) + (b_{ij}) = (a_{ij} + b_{ij})$$
$$\alpha(a_{ij}) = (\alpha a_{ij})$$
$$B = \{e_{ij}\}$$

where  $e_{ij}$  denotes a matrix of 0 everywhere except column i and row j.

(3) Functions with domain  $\Omega$ :

$$\mathcal{F}(\Omega, \mathbb{F}) = \{ f \mid f : \Omega \to \mathbb{F} \}$$
$$(f+g)(x) = f(x) + g(x)$$
$$(\alpha f)(x) = \alpha f(x)$$

(4) Bounded functions with domain  $\Omega$ :

$$\ell_{\infty}(\Omega, \mathbb{F}) = \{ f \in \mathcal{F}(\Omega, \mathbb{F}) \mid ||f||_{u} \le \infty \}$$
$$||f||_{u} = \sup_{x \in \Omega} |f(x)|$$

Exercises:

• Triangle Inequality:  $||f + g||_u \le ||f||_u + ||g||_u$ 

• Scalar Multiplication/Absolute Homogeneity:  $\|\alpha f\|_u = |\alpha| \|f\|_u$ 

• Positive Definite:  $||f||_u = 0 \Rightarrow f = 0$ 

**Proof of Triangle Inequality:** Given  $x \in \Omega$ ,

$$|(f+g)(x)| = |f(x) + g(x)|$$

$$\leq |f(x)| + |g(x)|$$

$$\leq ||f||_{u} + ||g||_{u}$$

Therefore.

$$\sup |(f+g)(x)| \le ||f||_u + ||g||_u$$
$$||f+g||_u \le ||f||_u + ||g||_u$$

(5) Continuous functions on closed and bounded intervals:

$$C([a, b], \mathbb{F}) = \{f : [a, b] \to \mathbb{F} \mid f \text{ continuous}\}\$$

Check that  $C([a, b], \mathbb{F}) \subseteq \ell_{\infty}([a, b], \mathbb{F})$  is a subspace.

(6) Let  $f : [a, b] \to \mathbb{R}$  be any function. Let  $\mathcal{P} : a = x_0 < x_1 < x_2 < \cdots < x_n = b$ .

$$\operatorname{var}(f; \mathcal{P}) := \sum_{k=1}^{n} |f(x_k) - f(x_{k-1})|$$

$$\operatorname{var}(f) = \sup_{\mathcal{P}} \operatorname{var}(f; \mathcal{P})$$

$$\operatorname{BV}([a, b]) = \{f : [a, b] \to \mathbb{R} \mid \operatorname{var}(f) < \infty\}$$

$$\|f\|_{\operatorname{BV}} = |f(a)| + \operatorname{var}(f)$$

BV([a, b]) is a vector space.

Question: Is  $\mathbb{1}_{\mathbb{Q}} \in BV([0,1])$ ?

(7) Suppose  $K \subseteq V$  is a *convex* subset of a vector space:  $v, w \in K, t \in [0, 1] \Rightarrow (1 - t)v + tw \in K$ . Let  $Aff(K) = \{f : K \to \mathbb{R} \mid f \text{ is affine}\}$ , where f is affine if  $\forall v, w \in K, t \in [0, 1], f((1 - t)v + tw) = (1 - t)f(v) + tf(w)$ .

**Exercise:** Show that  $Aff(K) \subseteq \mathcal{F}(K, \mathbb{R})$  is a subspace.

(8) Let S be defined as

$$S = \{(a_k)_{k=1}^{\infty} \mid a_k \in \mathbb{F}\}.$$

Under pointwise operations, S is a vector space.

$$(a_k)_k + (b_k)_k = (a_k + b_k)_k$$
$$\alpha(a_k)_k = (\alpha a_k)_k$$

Note 1:  $S = \mathcal{F}(\mathbb{N}, \mathbb{F})$ .

Note 2:  $c_{00} \subseteq \ell_1 \subseteq c_0 \subseteq c \subseteq \ell_\infty \subseteq S$ .

- $c_{00} = \{(a_k)_k \mid \text{finitely many } a_k \neq 0\}$
- $c_0 = \{(a_k)k \mid (a_k)_k \to 0\}$

- $c = \{(a_k)_k \mid (a_k)_k \to a < \infty\}$
- $\ell_{\infty} = \{(a_k)_k \mid ||(a_k)_k||_u < \infty\}$
- $\ell_1 = \{(a_k)_k \mid \sum_{k=1}^{\infty} |a_k| = a < \infty \}$
- (9)  $C_C(\mathbb{R}) \subseteq C_0(\mathbb{R}) \subseteq \ell_\infty(\mathbb{R})$  are all subspaces.
  - $C_C(\mathbb{R}) = \{f : \mathbb{R} \to \mathbb{F} \mid f \text{ compactly supported}\}: f : \mathbb{R} \to \mathbb{F} \text{ is compactly supported if } \exists [a, b] \text{ such that } x \notin [a, b] \Rightarrow f(x) = 0.$
  - $C_0(\mathbb{R}) = \{ f : \mathbb{R} \to \mathbb{F} \mid f \text{ continuous, } \lim_{x \to \pm \infty} f(x) = 0 \}$
- (10) Let S be any non-empty set.

$$\mathbb{F}(S) := \{ f : S \to \mathbb{F} \mid f \text{ finitely supported} \}$$
  
$$\mathsf{supp}(f) = \{ x \in S \mid f(x) \neq 0 \}$$

We claim that  $\mathbb{F}(S) \subseteq \mathcal{F}(S, \mathbb{F})$  is a subspace. Consider  $e_t : S \to \mathbb{F}$  defined as follows:

$$e_t(s) = \begin{cases} 1 & s = t \\ 0 & s \neq t \end{cases}.$$

We claim that  $\xi = \{e_t\}_{t \in S}$  is a basis for  $\mathbb{F}(S)$ .

Indeed, given  $f \in \mathbb{F}(S)$ , we know that  $\operatorname{supp}(f) = \{t_1, \ldots, t_n\} \subseteq S$ . Therefore,  $f = \sum_{k=1}^n f(t_k) e_{t_k} \in \operatorname{span}(\xi)$ . Therefore,  $\xi$  is spanning for  $\mathbb{F}(S)$ . Suppose  $\sum_{k=1}^n \alpha_{t_k} e_{t_k} = \emptyset$  for some  $\alpha_k \in \mathbb{F}$ ,  $t_k \in S$ .

$$\left(\sum_{k=1}^{lpha_{t_k}} e_{t_k}\right) = \mathbb{O}(t_1)$$
  $lpha_{t_1} = 0.$ 

Similarly,  $\alpha_{t_j} = 0$  for j = 1, ..., n. Therefore,  $\xi$  is linearly independent. Since  $\xi$  is linearly independent and spanning,  $\xi$  forms a basis for  $\mathbb{F}(S)$ .

**Note:** The free vector space,  $\mathbb{F}(S)$ , displays the universal property.

There are functions  $\iota: S \to \mathbb{F}(S)$ , where  $\iota(t) = e_t$ , and given any map  $\varphi: S \to V$  for V a vector space over  $\mathbb{F}$ ,  $\exists !$  linear map  $T_{\varphi}: \mathbb{F}(S) \to V$  such that  $\iota \circ T_{\varphi} = \varphi$ .

$$S \xrightarrow{\iota} \mathbb{F}(S)$$

$$\downarrow^{T_{\varphi}}$$

$$\downarrow^{V}$$

**Proof:** Every  $f \in \mathbb{F}(S)$  has a unique expression  $f = \sum_{k=1}^{n} f(t_k) e_{t_k}$ , where  $\text{supp}(f) = \{t_1, \dots, t_n\}$ . Therefore,

$$T_{\varphi}(f) := \sum_{k=1}^{n} f(t_k) \varphi(t_k)$$

**Exercise:** Show  $T_{\varphi}$  is linear and unique.

**Exercise 2:** Suppose V is a vector space over  $\mathbb{F}$  with basis B. Show that  $\mathbb{F}(B) \cong V$ . Remember that  $V \cong W$  if  $\exists \ T : V \to W$  such that T is bijective and linear.

### **Normed Spaces**

To every vector  $v \in V$ , we want to assign a length to v, ||v||.

A **norm** on a vector space V is a map

$$\|\cdot\|:V\to\mathbb{R}^+$$
$$v\mapsto\|v\|>0$$

such that

- (i) Homogeneity:  $\|\alpha v\| = |\alpha| \|v\|$
- (ii) Triangle Inequality:  $||v + w|| \le ||v|| + ||w||$

(iii) Positive definiteness:  $||v|| = 0 \Rightarrow v = \mathbb{O}_V$ .

If  $p: V \to \mathbb{R}^+$  satisfies (i) and (ii), then p is a **seminorm**.

The pair  $(V, \|\cdot\|)$  is called a normed space.

Two norms,  $\|\cdot\|$  and  $\|\cdot\|'$  are called **equivalent** if  $\exists c_1, c_2 \geq 0$  with,  $\forall v \in V$ ,

$$||v|| \le c_1 ||v||'$$
  
 $||v||' \le c_2 ||v||$ 

**Note:** On  $\mathbb{R}^n$ , all norms are equivalent.

**Exercise:** If p is any seminorm on V, then  $|p(v) - p(w)| \le p(v - w)$ .

**Notation:** If V is a normed space, then  $B_V = \{v \in V \mid ||v|| \le 1\}$ , and  $U_V = \{v \in V \mid ||v|| < 1\}$  are the closed and open unit ball respectively.

### **Examples of Normed Spaces**

(1) Given  $V = \mathbb{F}^n$  and  $X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$ , we have different norms:

$$\begin{split} \|x\|_1 &= \sum_{j=1}^n |x_j| \\ \|x\|_\infty &= \max_{1 \le j \le n} |x_j| \\ \|x\|_2 &= \left(\sum_{j=1}^n |x_j|^2\right)^{1/2}. \end{split}$$

In general, for  $1 \le p < \infty$ ,

$$||x||_p = \left(\sum_{j=1}^{\infty} |x_j|^p\right)^{1/p}.$$

**Exercise:** Show that  $\|\cdot\|_1$  and  $\|\cdot\|_\infty$  are norms. Show that  $\lim_{p\to\infty}\|x\|_p=\|x\|_\infty$ 

We want to show that  $\|\cdot\|_p$  defines a norm for  $1 \le p < \infty$ . If  $1 \le p < \infty$ , its conjugate index  $q \in [1, \infty]$  whereby  $\frac{1}{p} + \frac{1}{q} = 1$ . For example, if p = 1, then  $q = \infty$ , and if  $p = \infty$ , then q = 1.

**Lemma 1:** For  $1 , <math>p^{-1} + q^{-1} = 1$ ,  $f: [0, \infty) \to \mathbb{R}$ ,  $f(t) = \frac{1}{p}t^p - t + \frac{1}{q}$ . Then,  $f(t) \ge 0$  for all  $t \ge 0$ .

**Proof 1:** We can see that  $f'(t) = t^{p-1} - 1$ . Then, f'(t) = 0 at t = 1; f'(t) > 0 for t > 1 and f'(t) < 0 for  $t \in [0, 1)$ .

So, since  $f(t) \ge f(1)$  for all  $t \ge 0$ , and f(1) = 0,  $f(t) \ge 0$  for all  $t \ge 0$ .

**Lemma 2:** For  $1 , <math>p^{-1} + q^{-1} = 1$ ,  $z, y \ge 0$ ,  $xy \le \frac{1}{p}x^p + \frac{1}{q}y^q$ .

**Proof 2:** We know from Lemma 1,  $t \leq \frac{1}{p}t^p + \frac{1}{q}$ . Multiply by  $y^q$  to get

$$ty^q \le \frac{1}{p}t^p y^q + \frac{1}{q}y^q.$$

Set  $t = xy^{1-q}$ . Then,

$$xy^{1-q}y^q \le \frac{1}{p}x^py^{p-pq}y^q + \frac{1}{q}y^q.$$

Since  $\frac{1}{p} + \frac{1}{q} = 1$ , p - pq = -q, so

$$xy \leq \frac{1}{p}x^p + \frac{1}{q}y^q$$
.

With these two lemmas in mind, we get two important inequalities.

**Hölder's Inequality:** For  $1 \le p \le \infty$ ,  $p^{-1} + q^{-1} = 1$ . Then, for  $x, y \in \mathbb{F}^n$ ,

$$\left|\sum_{j=1}^n x_j y_j\right| \le \|x\|_p \|y\|_q.$$

**Proof of Hölder's Inequality:** For p = 1, the solution is as follows:

$$\left| \sum_{j=1}^{n} x_j y_j \right| \le \sum_{j=1}^{n} |x_j| |y_j|$$

$$\le \sum_{j=1}^{n} |x_j| ||y||_{\infty}$$

$$= ||x||_{\theta} ||y||_{\infty},$$

and similarly for  $p = \infty$ , q = 1.

For  $1 , assume <math>||x||_p = ||y||_q = 1$ .

$$\left| \sum_{j=1}^{n} x_{j} y_{j} \right| \leq \sum_{j=1}^{\infty} |x_{j}| |y_{j}|$$

$$\leq \sum_{j=1}^{n} \left( \frac{1}{p} |x_{j}|^{p} + \frac{1}{q} |y_{j}|^{q} \right)$$

$$= \frac{1}{p} \left( \sum_{j=1}^{n} |x_{j}|^{p} \right) + \frac{1}{q} \left( \sum_{j=1}^{n} |y_{j}|^{q} \right)$$

$$= \frac{1}{p} + \frac{1}{q}$$

$$= 1$$

If  $||x||_p = 0$  or  $||y||_q = 0$ , then  $x = \mathbb{O}_{\mathbb{F}}$  or  $y = \mathbb{O}_{\mathbb{F}}$ , the inequality still holds.

Assume  $||x||_p \neq 0$ ,  $||y||_p \neq 0$ . Set

$$x' = \frac{x}{\|x\|_{\rho}}$$
$$y' = \frac{y}{\|y\|_{\rho}}.$$

It can be verified that  $\|x'\|_p = 1 = \|y'\|_q$ . Therefore,

$$\left| \sum_{j=1}^{n} x_j' y_j' \right| \le 1$$

$$\left| \sum_{j=1}^{n} \frac{x_j}{\|x\|_p} \frac{y_j}{\|y\|_q} \right| \le 1$$

$$\left| \sum_{j=1}^{n} x_j y_j \right| \le \|x\|_p \|y\|_q$$

Minkowski's Inequality: Given  $x, y \in \mathbb{F}^n$ ,  $1 \le p \le \infty$ ,  $\frac{1}{p} = \frac{1}{q} = 1$ ,

$$||x + y||_p \le ||x||_p + ||y||_p$$

**Proof of Minkowski's Inequality:** We can verify for p = 1,  $q = \infty$ , and vice versa.

Assume 1 . Then,

$$\begin{split} \|x+y\|_{\rho}^{p} &= \sum_{j=1}^{n} |x_{j}+y_{j}|^{p} \\ &= \sum_{j=1}^{\infty} |x_{j}+y_{j}||x_{j}+y_{j}|^{p-1} \\ &\leq \sum_{j=1}^{\infty} |x_{j}||x_{j}+y_{j}|^{p-1} + \sum_{j=1}^{n} |y_{j}||x_{j}+y_{j}|^{p-1} \\ &\leq \left(\sum_{j=1}^{n} |x_{j}|^{p}\right)^{1/p} \left(\sum_{j=1}^{n} |x_{j}+y_{j}|^{pq-q}\right)^{1/q} + \left(\sum_{j=1}^{n} |y_{j}|^{p}\right)^{1/p} \left(\sum_{j=1}^{n} |x_{j}+y_{j}|^{pq-q}\right)^{1/q} \\ &= \|x\|_{\rho} \|x+y\|_{\rho}^{p/q} + \|y\|_{\rho} \|x+y\|_{\rho}^{p/q} \\ &= (\|x\|_{\rho} + \|y\|_{\rho}) \|x+y\|_{\rho}^{p-1} \end{split}$$

Divide by  $||x + y||_p^{p-1}$  to get desired inequality.

(2)  $\ell_{\infty}(\Omega, \mathbb{F})$  with  $\|\cdot\|_u$ . This includes subspaces that inherit the norm, such as

$$C([a, b]) \subseteq \ell_{\infty}(\Omega)$$
$$\ell_{\infty}(\mathbb{R}) \supseteq C_{0}(\mathbb{R}) \supseteq C_{C}(\mathbb{R})$$

**Exercise:** Show that  $C_0(\mathbb{R}) \subseteq \ell_\infty(\mathbb{R})$  is a subspace

(3)  $\Omega=\mathbb{N}$ ,  $\boldsymbol{\ell}_{\infty}=\boldsymbol{\ell}_{\infty}(\mathbb{N})$  with  $\|\cdot\|_{\infty}$ . Subspaces that inherit the norm are

$$c_{00} \subseteq c_0 \le \ell_{\infty}$$
.

(4)  $\ell_1$  with  $\|\cdot\|_1$ ,

$$||(a_k)_k||_1 = \sum_{k=1}^n |a_k|.$$

(5) C([a, b]) with

$$||f||_1 = \int_a^b |f(x)| dx.$$

(6) Let  $1 \le p < \infty$ .

$$\ell_p = \left\{ (a_k)_{k=1}^{\infty} \mid \sum_{k=1}^{\infty} |a_k|^p < \infty \right\}$$

is a normed space with

$$\|(a_k)_k\|_p = \left(\sum_{k=1}^{\infty} |a_k|^p\right)^{1/p}$$

We will show that the triangle inequality holds for this norm.

$$\left(\sum_{k=1}^{n} |a_k + b_k|^p\right)^{1/p} = \left\| \begin{bmatrix} a_1 + b_1 \\ \vdots \\ a_n + b_n \end{bmatrix} \right\|_{\ell_p^n}$$

$$= \left\| \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} + \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \right\|_{\ell_p^n}$$

$$\leq \left\| \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \right\| + \left\| \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} \right\|_{\ell_p^n}$$

$$\leq \left\| (a_k)_k \right\|_p + \left\| (b_k)_k \right\|_p.$$

Taking the limit as  $n \to \infty$  (by the definition of an infinite series), we find that  $\|(a_k)_k + (b_k)_k\|_p \le \|(a_k)_k\|_p + \|(b_k)_k\|_p$ .

(7)  $BV([a,b]) = \{f : [a,b] \to \mathbb{R} \mid Var(f) < \infty\}$  with the norm  $||f||_{BV} = |f(a)| + Var(f)$  is a normed space:

$$||f||_{BV} = 0$$
$$|f(a)| = 0$$
$$Var(f) = 0$$

given  $t \in (a, b]$ , look at the partition  $a < t \le b$ . Then,

$$Var(f) \ge |f(t) - f(a)| + |f(b) - f(t)|$$

$$f(t) = 0$$

$$f = \mathbb{O}_f.$$

(8)  $\mathbb{M}_{m,n}(\mathbb{F})$  with

$$||a||_{\text{op}} = \sup_{\|\xi\|_{\ell_2^n} \le 1} ||a\xi||_{\ell_2^m}$$

is a normed vector space. If  $||a||_{op} = 0$ , then

$$ae_j = 0$$
  $\forall j \in \{1, \dots, n\}.$ 

take the dot product with  $i \neq j$ 

$$ae_j \cdot e_i = a_{ij}$$
$$= 0$$

so  $a_{ij} = 0$  for all  $a_{ij}$ , so a is the 0 matrix.

(9) Let V, W be vector spaces over  $\mathbb{F}$ . Then,  $\mathcal{L}(V, W) = \{T \mid T : V \to W \text{ linear}\}$ , where  $T(\alpha v_1 + \beta v_2) = \alpha T(v_1) + \beta T(v_2)$ .

 $\mathcal{L}(V,W)$  is a vector space with operations

$$(T+S)(v) = T(v) + S(v)$$
$$(\alpha T)(v) = \alpha T(v).$$

**Notation:**  $\mathcal{L}(V) := \mathcal{L}(V, V)$  is all linear operators on V.  $\mathcal{L}(V, \mathbb{F}) = V'$  is all linear functionals.

Suppose V and W are normed vector spaces. If  $T: V \to W$ , set

$$||T||_{op} := \sup_{\|v\|_{V} \le 1} ||T(v)||_{W},$$

$$\mathbb{B}(V, W) = \{T \in \mathcal{L}(V, W) \mid ||T||_{op} \le \infty\},$$

where  $\mathbb{B}(V,W)$  is referred to as the set of all bounded linear maps from V to W.  $\mathbb{B}(V,W)$  with  $\|\cdot\|_{\mathrm{op}}$  is a normed space.

• Homogeneity:

$$\begin{split} \|\alpha T\|_{[op]} &= \sup_{\|v\|_{V} \le 1} \|\alpha T(v)\|_{W} \\ &= \sup_{\|v\|_{V} \le 1} |\alpha| \|T(v)\|_{W} \\ &= |\alpha| \sup_{\|v\|_{V} \le 1} \|T(v)\|_{W} \\ &= |\alpha| \|T\|_{\text{op}}. \end{split}$$

• Triangle Inequality: for  $||v||_V \le 1$ ,

$$|| (T+S) (v) ||_{W} = || T(v) + S(v) ||_{W}$$

$$\leq || T(v) ||_{W} + || S(v) ||_{W}$$

$$\leq || T ||_{op} + || S ||_{op}$$

so

$$||T + S||_{op} = \sup_{||v|| \le 1} ||T + S(v)||$$
  
  $\le ||T||_{op} + ||S||_{op}$ 

• Positive Definite: If  $||T||_{op} = 0$ , then T(v) = 0 for all  $v \in V$ ,  $||v|| \le 1$ .

Let  $v \in V$ ,  $v \neq 0$ . Then,  $\frac{v}{\|v\|} \in B_V$ .

$$T\left(\frac{v}{\|v\|}\right) = 0$$

$$\frac{1}{\|v\|}T(v) = 0$$

$$T(v) = 0$$

Special Cases:  $\mathbb{B}(V) = \mathbb{B}(V, V), V^* = \mathbb{B}(V, \mathbb{F}).$ 

Exercise:  $\mathcal{L}(\mathbb{F}^n, \mathbb{F}^m) = \mathbb{B}(\ell_2^n, \ell_2^m)$ .

(10) Inner Product Spaces (expanded upon below).

#### Inner Product Spaces

An inner product on a vector space V is a pairing

$$V \times V \xrightarrow{\langle \cdot, \cdot \rangle} \mathbb{F}$$

that satisfies

- (i)  $\langle v_1 + v_2, w \rangle = \langle v_1, w \rangle + \langle v_2, w \rangle$ ,  $\langle \alpha v, w \rangle = \alpha \langle v, w \rangle$ .
- (ii)  $\langle v, w \rangle = \overline{\langle w, v \rangle}$
- (iii)  $\langle v, v \rangle \geq 0$ .
- (iv) If  $\langle v, v \rangle = 0$ , then v = 0.

The pair  $(V, \langle \cdot, \cdot \rangle)$  is known as an inner product space.

Remarks:  $\langle v, w_1 + w_2 \rangle = \langle v, w_1 \rangle + \langle v, w_2 \rangle, \langle v, \alpha w \rangle = \overline{\alpha} \langle v, w \rangle.$ 

If  $\langle \cdot, \cdot \rangle$  is an inner product on a linear space V, then set

$$||v||_2 := \langle v, v \rangle^{1/2}.$$

Exercise:  $\|\alpha v\|_2 = |\alpha| \|v_2\|, \|v\|_2 = 0 \Rightarrow v = 0.$ 

 $v, w \in (V, \langle, \cdot, \cdot\rangle)$  are orthogonal if  $\langle v, w \rangle = 0$ .

The Pythagoran theorem states that for  $v_1, \ldots, v_n \in V$  mutually orthogonal, then

$$\left\| \sum_{i=1}^{n} v_i \right\|^2 = \sum_{i=1}^{n} \|v_i\|^2.$$

For two vectors  $v, w \in V$ ,  $P_w(v) = \frac{\langle v, w \rangle}{\langle w, w \rangle} w$ .

**Exercise:** Check that  $\langle P_w(v), v - P_w(v) \rangle$ , meaning

$$||v||^2 = ||P_w(v)||^2 + ||v - P_w(v)||^2$$

Cauchy-Schwarz Inequality: In any inner product space,

$$|\langle v, w \rangle| \leq ||v|| \cdot ||w||$$
.

Proof of Cauchy-Schwarz: From the exercise,

$$||v|| \ge ||P_w(v)||$$

$$||v|| \ge \left\| \frac{\langle v, w \rangle}{\langle w, w \rangle} w \right\|$$

$$= \frac{|\langle v, w \rangle|}{||w||^2} ||w||$$

therefore,

$$||v||||w|| \ge |\langle v, w \rangle|$$

The triangle inequality follows from the Cauchy-Schwarz inequality.

## Proof of Triangle Inequality:

$$||v + w||_{2}^{2} = \langle v + w, v + w \rangle$$

$$= \langle v, v \rangle + \langle v, w \rangle + \langle w, v \rangle + \langle w, w \rangle$$

$$= ||v||^{2} + ||w||^{2} + \langle v, w \rangle + \overline{\langle v, w \rangle}$$

$$= ||v||^{2} + ||w||^{2} + 2\operatorname{Re}\langle v, w \rangle$$

$$\leq ||v||^{2} + ||w||^{2} + 2|\langle v, w \rangle|$$

$$\leq ||v||^{2} + ||w||^{2} + 2||v|||w||$$

$$= (||v|| + ||w||)^{2}.$$

Cauchy-Schwarz Inequality

Take square roots on both sides.

(1)  $\ell_2^n = \mathbb{F}^n$  with

$$\left\langle \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}, \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \right\rangle = \sum_{i=1}^n x_i \overline{y_i}.$$

Cauchy-Schwarz is found as

$$\left| \sum_{j=1}^{n} x_{j} \overline{y_{j}} \right| \leq \left( \sum_{j=1}^{n} |x_{j}|^{2} \right)^{1/2} \left( \sum_{j=1}^{n} |y_{j}|^{2} \right)^{1/2}.$$

(2)  $\ell_2$  with

$$\langle (a_j)_j, (b_j)_j \rangle = \sum_{j=1}^{\infty} a_j \overline{b}_j.$$

We can see that for any finite n, the Cauchy-Schwarz inequality in  $\ell_2^n$  states

$$\begin{split} \left| \sum_{j=1}^{n} a_{j} \overline{b_{j}} \right| &\leq \left( \sum_{j=1}^{n} |a_{j}|^{2} \right)^{1/2} \left( \sum_{j=1}^{n} |b_{j}|^{2} \right)^{1/2} \\ &\leq \left( \sum_{j=1}^{\infty} |a_{j}|^{2} \right)^{1/2} \left( \sum_{j=1}^{\infty} |b_{j}|^{2} \right)^{1/2}. \end{split}$$

Taking the limit as  $n \to \infty$ , we see that  $\langle (a_j)_j, (b_j)_j \rangle$  is convergent.

(3) C([a, b]) with

$$\langle f, g \rangle = \int_{a}^{b} f(x) \overline{g(x)} dx.$$

(4) Let  $V = \mathbb{M}_n(\mathbb{C})$ .

Recall that if

$$a=(a_{ij})_{i,j},$$

then

$$a^* = (\overline{a_{ii}})_{i,j}$$
.

Let  $\operatorname{Tr}: \mathbb{M}_n(\mathbb{C}) \to \mathbb{C}$ ,  $\operatorname{Tr}((a_{ij})) = \sum_{i=1}^n a_{ii}$ .

- $Tr(I_n) = n$
- $Tr(a + \alpha b) = Tr(a) + \alpha Tr(b)$
- Tr(ab) = Tr(ba)

Then, if  $Tr(a^*a) = 0$ , then  $a = \mathbb{O}_{M_n}$ .

$$a^*a = (\overline{a_{ji}})_{i,j}(a_{ij})_{i,j}$$

$$= \left(\sum_{k=1}^n \overline{ki} a_{kj}\right)_{i,j}$$

$$\operatorname{Tr}(a^*a) = \sum_{i=1}^n \sum_{k=1}^n \overline{a_{ki}} a_{ki}$$

$$= \sum_{i,k=1}^n |a_{ki}|^2$$

$$= \sum_{i,j=1}^n |a_{ij}|^2.$$

If  $Tr(a^*a) = 0$ , then  $a_{ij} = 0$  for all i, j.

We define

$$\langle a, b \rangle_{\mathsf{HS}} = \mathsf{Tr}(b^*a).$$

(i) 
$$(b_1 + b_2)^* = b_1^* + b_2^*$$

(ii) 
$$(\alpha b)^* = \overline{\alpha} b^*$$

(iii) 
$$(b_1b_2)^* = b_2^*b_1^*$$

(iv) 
$$b^{**} = b$$

The norm is defined as

$$||a||_{HS} = \langle a, a \rangle^{1/2}$$
  
=  $Tr(a^*a)^{1/2}$   
=  $\left(\sum_{i,j=1}^n |a_{ij}|^2\right)^{1/2}$ 

# **Metric Spaces**

We looked at normed spaces, where we attach a length  $\|v\|$  to very vector v. We can also speak of the distance between two vectors, defined as  $d(v, w) = \|v - w\|$ .

Notice that the following hold:

• 
$$d(v, w) \geq 0$$

•

$$d(v, w) = ||v - w||$$

$$= ||(-1)(w - v)||$$

$$= |-1||w - v||$$

$$= ||w - v||$$

•

$$d(u, w) = ||u - w||$$

$$= ||u - v + v - w||$$

$$\leq ||u - v|| + ||v - w||$$

$$= d(u, v) + d(v, w).$$

• d(v, v) = ||v - v|| = 0. If d(v, w) = 0, then ||v - w|| = 0, so v - w = 0, so v = w.

In Real Analysis I, we studied the properties (such as convergence, limits, and continuity) of a particular normed vector space, namely  $(\mathbb{R}, |\cdot|)$ . We will expand these concepts to all metric spaces.

## **Definition of a Metric Space**

Let X be a non-empty set. A **metric** on X is a map

$$d: X \times X \to \mathbb{R}^+$$
$$(x, y) \mapsto d(x, y) \ge 0$$

such that

- (i) Symmetry: d(x, y) = d(y, x) for all  $x, y \in X$ .
- (ii) Triangle Inequality:  $d(x, z) \le d(x, y) + d(y, z)$  for all  $x, y, z \in X$ .
- (iii) Zero Distance: d(x, x) = 0
- (iv) Definite:  $d(x, y) = 0 \Rightarrow x = y$

If d satisfies (i), (ii), and (iii), then d is called a semi-metric. If d satisfies (iv) as well, then d is a metric.

If d is a (semi-)metric on X, the pair (X, d) is called a (semi-)metric space.

Two metrics, d and  $\rho$ , on X, are equivalent if  $\exists c_1, c_2 \geq 0$  such that  $d(x, y) \leq c_1 \rho(x, y)$  and  $\rho(x, y) \leq c_2 d(x, y)$  for all x, y.

### **Examples of Metric Spaces**

(1) Discrete Metric:

$$d(x,y) = \begin{cases} 1 & x \neq y \\ 0 & x = y \end{cases}$$

for X any set.

(2) Hamming distance: between two bit strings of equal length. Let

$$X = \{0, 1\}^n$$

$$= \{0, 1\} \underbrace{\times \cdots \times}_{n \text{ times}} \{0, 1\}$$

$$d_H((x_i)_1^n, (y_i)_1^n) = |\{j \mid x_i \neq y_i\}|.$$

(3) Any normed space  $(V, \|\cdot\|)$  is a metric space.

$$d(v,w) = ||v-w||.$$

Exercise: Show that if two norms are equivalent, their induced metrics are equivalent.

- (4) Subset of Metric Space: If (X, d) is a metric space, and  $Y \subseteq X$  is non-empty. Then, (Y, d) is a metric space.
- (5) Paris metric: let  $(X, \rho)$  be a metric space. Let  $p \in X$  be a fixed point.

$$\rho(x,y) := \begin{cases} 0 & x = y \\ \rho(x,p) + \rho(p,y) & x \neq y \end{cases}$$

(6) Bounded metric: Let  $\rho$  be a (semi-)metric on X. Set

$$d(x,y) = \frac{\rho(x,y)}{1 + \rho(x,y)}.$$

We claim that d is a (semi-)metric. Notice that  $0 \le d(x, y) \le 1$ .

**Proof:** Clearly, d(x, y) = d(y, x). Additionally, d(x, x) = 0. If d(x, y) = 0 and  $\rho$  is a metric, then  $\rho(x, y) = 0$ , so x = y.

To show the triangle inequality, we examine the function

$$f(t) = \frac{t}{1+t}$$
$$f'(t) = \frac{1}{(1+t)^2} > 0.$$

Since  $\rho$  satisfies the triangle inequality,  $\rho(x,z) \le \rho(x,y) + \rho(y,z)$ . Apply f on both sides. Then,

$$\underbrace{\frac{\rho(x,z)}{1+\rho(x,z)}}_{d(x,z)} \le \frac{\rho(x,y)+\rho(y,z)}{1+(\rho(x,y)+\rho(y,z))} 
= \frac{\rho(x,y)}{1+\rho(x,y)+\rho(y,z)} + \frac{\rho(y,z)}{1+\rho(x,y)+\rho(y,z)} 
\le \underbrace{\frac{\rho(x,y)}{1+\rho(x,y)}}_{d(x,y)} + \underbrace{\frac{\rho(y,z)}{1+\rho(y,z)}}_{d(y,z)}.$$

(7) If  $d_1, \ldots, d_n$  are metrics on  $X, c_1, \ldots, c_n \ge 0$ . Then,

$$d(x,y) = \sum_{k=1}^{n} c_k d_k(x,y)$$

is a metric.

(8) Let  $\{\rho_k\}_{k=1}^{\infty}$  be a family of semi-metrics. Assume the family is separating — for all  $x \neq y$ , there exists k such that  $\rho_k(x,y) \neq 0$ .

Let  $d_k$  be defined as

$$d_k(x,y) = \frac{\rho_k(x,y)}{1 + \rho_k(x,y)}.$$

Note that  $\{d_k\}_{k=1}^{\infty}$  is also separating.

Then,

$$d(x,y) = \sum_{k=1}^{\infty} 2^{-k} d_k(x,y)$$

is a metric.

We will now define the Frechet Metric using this method. Let  $X=C(\mathbb{R})$ . For each  $k=1,2,3,\ldots$ , set  $p_k(f)=\sup_{x\in [-k,k]}|f(x)|$ .

We can verify that  $p_k$  defines a seminorm. We can then check  $\rho_k(f,g)=p_k(f-g)$  is a semi-metric.

We claim that  $\{\rho_k\}$  is separating: if  $f \neq g$ , then there exists  $x_0 \in \mathbb{R}$  with  $f(x_0) \neq g(x_0)$ . Since f and g are continuous, there is a neighborhood  $[x_0 - \delta, x_0 + \delta]$  such that  $f(x) \neq g(x)$  for all  $x \in [x_0 - \delta, x_0 + \delta]$ . Find k such that  $[x_0 - \delta, x_0 + \delta] \subseteq [-k, k]$ . Then,  $\rho_k(f - g) > 0$ .

Construct  $d_k$  as above, and then d as follows:

$$d_{\mathsf{F}} = \sum \frac{2^{-k} p_k(f - g)}{1 + p_k(f - g)}$$

(9) Product of metric spaces: let  $(X_k, \rho_k)_{k=1}^{\infty}$  be a countable family of metric spaces. For each k, let

$$d_k(x,y) = \frac{\rho_k(x,y)}{1 + \rho_k(x,y)}.$$

**Remark:** If the  $\rho_k$  are already uniformly bounded, let  $d_k = \rho_k$ .

Let

$$X = \prod_{k=1}^{\infty} X_k$$

$$= \{ (x_k)_k \mid x_k \in X_k \}$$

$$= \left\{ f : \mathbb{N} \to \bigsqcup_{k=1}^{\infty} X_k \mid f(k) \in X_k \right\}.$$

Define  $D: X \times X \to [0, \infty)$  as

$$D(x, y) = \sum_{k=1}^{\infty} 2^{-k} \rho_k(x_k, y_k),$$

$$D(f, g) = \sum_{k=1}^{\infty} 2^{-k} \rho(f(k), g(k)).$$

For example, for each k, let  $X_k = \{0, 1\}$  with the discrete metric. Let

$$\Delta = \prod_{k \in \mathbb{N}} \{0, 1\}$$

$$= \{(x_k)_k \mid x_k \in \{0, 1\}\}$$

$$D(x, y) = \sum_{k=1}^{\infty} 2^{-k} |x_k - y_k| \qquad (x_k)_k, (y_k)_k \in \Delta.$$

 $\Delta$  is known as the abstract Cantor set; every compact metric space is a surjective image of the abstract Cantor set.

(10) Geodesic Distance: let  $\langle \cdot, \cdot \rangle$  be the standard dot product on  $\mathbb{R}^3(\mathbb{R}^n)$ , then

$$S^{2} = \left\{ x \in \mathbb{R}^{3} \mid ||x||_{2} = 1 \right\}$$
$$S^{n-1} = \left\{ x \in \mathbb{R}^{n} \mid ||x||_{2} = 1 \right\}.$$

To find the geodesic distance, we take  $d(x, y) = \arccos(\langle x, y \rangle)$ . We claim d is a metric.

- Symmetry: self-evident.
- $d(x, x) = \arccos(1) = 0$ . Suppose d(x, y) = 0. Then,  $\langle x, y \rangle = 1$ , meaning  $||x y||^2 = 0$ , so x = y.
- Let  $\theta = \arccos(\langle x, y \rangle)$ ,  $\varphi = \arccos(\langle y, z \rangle)$ , where  $\theta, \varphi \in [0, \pi]$ .

$$p_{X} = \frac{\langle x, y \rangle}{\langle y, y \rangle} y$$
$$= \cos(\theta) y$$
$$x = \cos(\theta) y + \sin(\theta) u$$

where

$$u = \frac{x - p_X}{\|x - p_X\|}.$$

Similarly, we can take

$$z = \cos(\varphi)y + \sin(\varphi)v$$

where

$$v = \frac{z - p_z}{\|z - p_z\|}.$$

So,

$$\begin{split} \langle x,z\rangle &= \cos(\theta)\cos(\varphi) + \sin(\theta)\sin(\varphi)\,\langle u,v\rangle \\ &\geq \cos(\theta)\cos(\varphi) - \sin(\theta)\sin(\varphi) & \langle u,v\rangle \geq -1 \\ &= \cos(\theta+\varphi). \end{split}$$

Since arccos is decreasing,

$$\begin{aligned} \arccos(\langle x, z \rangle) &\leq \arccos(\cos(\theta + \varphi)) \\ &= \theta + \varphi \\ &= \arccos(\langle x, y \rangle) + \arccos(\langle y, z \rangle). \end{aligned}$$

Therefore,  $d(x, y) \le d(x, y) + d(y, z)$ .

• Let  $\Gamma = (V, E)$  be a simple connected graph. We define  $d: V \times V \to [0, \infty)$  to be the length of the shortest path between vertices u and v.

Exercise: Show this is a metric.

(11) Let (X, d) be any metric space. If  $E \subseteq X$ , define  $\operatorname{diam}(E) = \sup_{x,y \in E} d(x,y)$ . E is bounded if  $\operatorname{diam}(E) < \infty$ .

**Exercise:** If  $(V, \|\cdot\|)$  is a normed space,  $E \subseteq V$  is a subset, show the following are equivalent:

- (i) E is bounded (in the metric sense)
- (ii)  $\sup_{v \in E} \|v\| < \infty$
- (iii)  $\exists r > 0$  such that  $E \subseteq rB_V$ .

Let  $\Omega$  be any set. The function  $f:\Omega\to X$  is bounded if  $f(\Omega)\subseteq X$  is bounded. We let  $\mathrm{Bd}(\Omega,X)=\{f:\Omega\to X\mid f\text{ is bounded}\}$ .

Remark:  $Bd(\Omega, \mathbb{F}) = \ell_{\infty}(\Omega, \mathbb{F}).$ 

(12)  $Bd(\Omega, X)$  with

$$D_u(f,g) = \sup_{x \in \Omega} d(f(x), g(x)).$$

**Exercise:** Show that  $D_u$  defines a metric.

Consider  $Bd(\Omega, \mathbb{F}) = \ell_{\infty}$ . Look at the subset

$$E = \{ f \in Bd(\Omega, \mathbb{F}) \mid f(x) \in \{0, 1\} \}.$$

Then,

$$D_u(f, g) = \sup_{x \in \Omega} |f(x) - g(x)|.$$

$$= \begin{cases} 1 & f \neq g \\ 0 & f = g \end{cases}.$$

When we take a particular subset of  $D_u(f, g)$ , we find that we get the discrete metric.

Taking an overview of the concepts we have learned so far, we see

Inner Product Spaces  $\subseteq$  Normed Vector Spaces  $\subseteq$  Metric Spaces

# **Topology of Metric Spaces**

Throughout this section, let (X, d) be a metric space.

- (1) Let  $x_0 \in X$ ,  $\delta > 0$ .
  - (i) We say

$$U(x_0, \delta) = \{x \in X \mid d(x, x_0) < \delta\}$$

is the open ball centered at  $x_0$  with radius  $\delta$ .

(ii) We say

$$B(x_0, \delta) = \{x \in X \mid d(x, x_0) \le \delta\}$$

is the closed ball.

(iii) We say

$$S(x_0, \delta) = \{x \in X \mid d(x, x_0) = \delta\}$$

is the sphere.

(2)  $U \subseteq X$  is open if

$$(\forall x \in U)(\exists \delta > 0) \ni U(x, \delta) \subseteq U.$$

Let

$$\tau_X = \{ U \subseteq X \mid U \text{ open} \}$$
$$\subseteq \mathcal{P}(X).$$

(3)  $D \subseteq X$  is closed if  $D^c$  is open.

(4) If  $x \in U \in \tau_X$ , then U is called an open neighborhood of x. If  $x \in U \subseteq N$ , where  $U \in \tau_X$ , then N is a neighborhood of x.

$$\mathcal{N}_{x} = \{ N \mid N \text{ is a neighborhood of } x \}$$

(5) Let  $A \subseteq X$ . The interior of A is

$$A^0 = \bigcup \{ V \mid V \subseteq A, V \text{ open} \}$$
.

The closure of A is

$$\overline{A} = \bigcap \{D \mid A \subseteq D, D \text{ closed}\}.$$

The boundary of A is

$$\partial A = \overline{A} \setminus A^0$$
.

Exercise:  $\overline{A^c} = (A^0)^c$ ,  $(\overline{A})^c = (A^c)^0$ .

**Remarks:**  $A^0$  is the largest open set contained in A. So, if V is open and  $V \subseteq A$ , then  $V \subseteq A^0$ . Similarly,  $\overline{D}$  is the smallest closed set containing D. If C is closed and  $D \subseteq C$ , then  $\overline{D} \subseteq C$ .

- For example,  $(a, b]^0 = (a, b)$ . This is because (a, b) is open and contained in (a, b], so  $(a, b) \subseteq (a, b]^0$ .
- We will show that  $\overline{A^c} \subseteq (A^0)^c$ .

$$A^0 \subseteq A$$
$$(A^0)^c \supset A^c$$

The union of open sets is open, so  $A^0$  is open, so  $(A^0)^c$  is closed by definition. Therefore,

$$(A^0)^c \supset \overline{A^c}$$
.

## Topology of Open Sets in a Metric Space

The open sets  $\tau_X$  form a topology:

- (i)  $\emptyset$ ,  $X \in \tau_X$ .
- (ii) If  $\{V_i\}_{i\in I}\subseteq \tau_X$ , then

$$\bigcup_{i\in I}V_i\in\tau_X.$$

(iii) If  $V_i, \ldots, V_n \in \tau_X$ , then

$$\bigcap_{i=1}^n V_i \in \tau_X.$$

**Remark:** This is only true of finite intersections. For a counterexample, if  $V_n = (-1/n, 1/n) \subseteq \mathbb{R}$  with the Euclidean metric, then the infinite intersection yields  $\{0\}$ , which is closed in  $\mathbb{R}$  with the Euclidean metric.

Proof:

- (1) Clearly,  $\emptyset$  (by vacuous truth) and X are open.
- (2) Let  $x \in \bigcup_{i \in I} V_i$ . Then,  $\exists i_0 \in I$  with  $x \in V_{i_0}$ . Since  $V_{i_0}$  is open,  $\exists \varepsilon > 0$  such that  $U(x, \varepsilon) \subseteq V_{i_0} \subseteq \bigcup V_i$ .
- (3) Let  $x \in \bigcap_{i=1}^n V_i$ . Then,  $x \in V_i$  for all  $i \in 1, ..., n$ . Since each  $V_i$  is open,  $\exists \varepsilon_1, ..., \varepsilon_n$  with  $U(x, \varepsilon_i) \subseteq V_i$  for each i = 1, ..., n. Set  $\varepsilon = \min\{\varepsilon_i\}_{i=1}^n$ . Then,  $U(x, \varepsilon) \subseteq U(x, \varepsilon_i) \subseteq V_i$  for all i. Therefore,  $U(x, \varepsilon) \subseteq \bigcap_{i=1}^n V_i$ .

Exercise: Show all open balls are open. In particular, show all open intervals are open.

Exercise: Show the following:

- (1)  $X, \emptyset$  are closed.
- (2) If  $\{C_i\}_{i\in I}$  is a family of closed sets, then  $\bigcap_{i\in I} C_i$  is closed.
- (3) For  $C_1, \ldots, C_n$  closed, then  $\bigcup_{i=1}^n C_i$  is closed.
- (4) Closed balls are closed. Spheres are closed.

Let  $x \in X$ . Recall that  $\mathcal{N}_X$  is the set of all neighborhoods of x.

- (i)  $N \in \mathcal{N}_X \Leftrightarrow \exists \delta > 0 : U(x, \delta) \in N$
- (ii)  $N \in \mathcal{N}_X$ ,  $N \subseteq M \Rightarrow M \in \mathcal{N}_X$
- (iii)  $N_1$ ,  $N_2 \in \mathcal{N}_x \Rightarrow N_1 \cap N_2 \in \mathcal{N}_x$

In this sense,  $\mathcal{N}_{x}$  is a directed set with reverse inclusion.

## **Pointwise Characterization of Subsets**

Let  $A \subseteq X$ .

(i)  $x \in A^0 \Leftrightarrow \exists \delta > 0 : U(x, \delta) \subseteq A$ .

(ii)  $x \in \overline{A} \Leftrightarrow \forall \delta > 0 : U(x, \delta) \cap A \neq \emptyset$ .

(iii)  $x \in \partial A \Leftrightarrow \forall \delta > 0 : U(x, \delta) \cap A \neq \emptyset$  and  $U(x, \delta) \cap A^c \neq \emptyset$ .

**Proof:** Let  $A \subseteq X$ 

(i)

$$x \in A^{0} \Leftrightarrow x \in \bigcup_{\substack{V \in \tau_{X} \\ V \subseteq A}} V$$
$$\Leftrightarrow \exists V \in \tau_{X}, V \subseteq A, x \in V$$
$$\Leftrightarrow \exists \delta > 0 : U(x, \delta) \subseteq A.$$

(ii)

$$x \notin \overline{A} \Leftrightarrow x \in (\overline{A})^{c}$$

$$\Leftrightarrow x \in (A^{c})^{0}$$

$$\Leftrightarrow \exists \delta > 0 : U(x, \delta) \subseteq A^{c}$$

$$\Leftrightarrow \exists \delta > 0 : U(x, \delta) \cap A = \emptyset.$$

We negate both sides.

(iii)

$$x \in \partial A \Leftrightarrow x \in \overline{A} \setminus A^{0}$$

$$\Leftrightarrow x \in \overline{A} \cap (A^{0})^{c}$$

$$\Leftrightarrow x \in \overline{A} \cap \overline{A}^{c}$$

$$\Leftrightarrow x \in \overline{A} \text{ and } x \in \overline{A}^{c}$$

$$\Leftrightarrow \forall \delta > 0 : U(x, \delta) \cap A \neq \emptyset, U(x, \delta) \cap A^{c} \neq \emptyset$$

**Proof:** We show that  $\overline{U}(v,\delta) = B(v,\delta)$ . Since  $B(v,\delta)$  is closed, and  $U(v,\delta) \subseteq B(v,\delta)$ , we know  $\overline{U(v,\delta)} \subseteq B(v,\delta)$ .

Let  $w \in B(v, \delta)$ . If  $||w - v|| < \delta$ , then  $w \in U(v, \delta)$ . Assume  $||w - v|| = \delta$ . Let  $u_t = (1 - t)v + tw$ , where  $t \in [0, 1]$ .

$$||w - u_t|| = ||w - (1 - t)v - tw||$$

$$= ||(1 - t)(w - v)||$$

$$= (1 - t)||w - v||$$

$$= (1 - t)\delta.$$

Let  $\varepsilon > 0$ . Let  $t \in (0,1)$  such that  $(1-t)\delta < \varepsilon$ . Then,  $u_t \in U(w,\varepsilon) \cap U(v,\delta)$ . Therefore,  $w \in \overline{U(v,\delta)}$ .

### Unions and Intersections of Closure/Interior

Let (X, d) be a metric space.

(i)

$$\left(\bigcup_{i\in I}A_i\right)^0\supseteq\bigcup_{i\in I}A_i^0$$

may be strict

(ii)

$$\overline{\bigcap_{i\in I} A_i} \subseteq \bigcap_{i\in I} \overline{A_i}$$

(iii)

$$\bigcap_{k=1}^{n} A_k^0 = \left(\bigcap_{k=1}^{n} A_k\right)^0$$

(iv)

$$\overline{\bigcup_{k=1}^{n} D_k} = \bigcup_{k=1}^{n} \overline{D_k}$$

Proof:

(i)

$$A_{i}^{0} \subseteq A_{i}$$

$$\bigcup_{i \in I} A_{i}^{0} \subseteq \bigcup_{i \in I} A_{i}$$

$$\bigcup_{i \in I} A_{i}^{0} \subseteq \left(\bigcup_{i \in I} A_{i}\right)^{0}$$

**Remark:** We claim  $\overline{\mathbb{Q}} = \mathbb{R}$  under the absolute value metric. We know that  $\mathbb{Q} \subseteq \mathbb{R}$ ,  $\mathbb{R}$  is closed, meaning  $\overline{\mathbb{Q}} \subseteq \mathbb{R}$ . Let  $t \in \mathbb{R}$ ,  $\delta > 0$ . We know that  $(t - \delta, t + \delta) \cap \mathbb{Q} \neq \emptyset$ . Therefore,  $t \in \overline{\mathbb{Q}}$ . Thus,  $\overline{\mathbb{Q}} = \mathbb{R}$ .

# **Properties of Boundary**

Let  $A \subseteq X$ .

- (1)  $\partial A$  is closed.
- (2)  $\partial A = \partial A^{c}$
- (3)  $\overline{A} = A \cup \partial A$
- (4)  $A \setminus \partial A = A^0$

Proof:

(1)

$$\partial A = \overline{A} \setminus A^0$$
$$= \overline{A} \cap (A^0)^{c}.$$

- (2) Follows from pointwise characterization.
- (3) Clearly,  $A \cup \partial A \subseteq \overline{A}$ . Let  $x \in \overline{A}$ . If  $x \in A$ , we're done. Otherwise,  $x \in \overline{A} \setminus A \subseteq \overline{A} \setminus A^0 = \partial A$ .
- (4)

$$A \setminus \partial A = A \cap (\partial A)^{c}$$

$$= A \cap (\overline{A} \setminus A^{0})^{c}$$

$$= A \cap (\overline{A} \cap (A^{0})^{c})^{c}$$

$$= A \cap (\overline{A}^{c} \cup A^{0})$$

$$= (A \cap \overline{A}^{c}) \cup (A \cap A^{0})$$

$$= A^{0}$$

## **Density and Separability**

Let (X, d) be a metric space.

- (1)  $A \subseteq X$  is d-dense if  $\overline{A} = X$ .
- (2)  $N \subseteq X$  is nowhere dense if  $(\overline{N})^0 = \emptyset$ .
- (3) (X, d) is separable if there is a countable dense subset.

**Exercise:** If  $N \subseteq X$  is closed, then N is nowhere dense if and only if  $N^c$  is dense.

Exercise: The following are equivalent.

- (1)  $A \subseteq X$  is dense.
- (2)  $\forall \emptyset \neq U \in \tau_X$ ,  $U \cap A \neq \emptyset$ .
- (3)  $\forall x \in X, \forall \varepsilon > 0, U(x, \varepsilon) \cap A \neq \emptyset$
- (4)  $\forall x \in X, \forall \varepsilon > 0, \exists a \in A \text{ such that } d(x, a) < \varepsilon.$

Let X be a metric space.

(1) A base for  $\tau_X$  is a family of open subsets  $\mathcal B$  such that:

$$(\forall U \in \tau_X) (\forall x \in U) \exists B \in \mathcal{B} \ni x \in B \subseteq U.$$

Equivalently,

$$\forall U \in \tau_X, U = \bigcup_{i \in I} B_i.$$
  $B_i \in \mathcal{B}$ 

- (2) We say that (X, d) is second countable if  $\tau_X$  admits a countable base.
- For any (X,d) a metric space,  $\mathcal{B} = \{U(x,\varepsilon) \mid x \in X, \varepsilon > 0\}$  is a base. Indeed, given any  $x \in U \subseteq \tau_X$ , by definition,  $\exists \varepsilon > 0$  such that  $U(x,\varepsilon) \subseteq U$ . Alternatively,  $\mathcal{B}' = \{U(x,1/n) \mid x \in X, n \geq 1\}$  is a topological base.
- Let  $X = \mathbb{R}^d$  with the Euclidean metric. Then, for  $\mathcal{B} = \{U(q, 1/n) \mid n \geq 1, q \in \mathbb{Q}^d\}$ , we claim this is a base.

Let  $V \subseteq \mathbb{R}^d$  be open,  $r \in V$ . Since V is open,  $\exists \delta > 0$  with  $U(r, \delta) \subseteq V$ . Find n large such that  $1/n < \delta$ . Find  $q \in \mathbb{Q}^d$  with ||r - q|| < 1/2n. This is always possible as  $\mathbb{Q}^d$  is dense in  $\mathbb{R}^d$ .

Consider U(q, 1/2n). Then,  $r \subseteq U(q, 1/2n) \subseteq U(r, \delta) \subseteq V$  because ||r - q|| < 1/2n, and if  $t \in U(q, 1/2n)$ , then

$$||t - r|| \le ||t - q|| + ||q - r||$$
  
 $< 1/2n + 1/2n$   
 $= 1/n$   
 $< \delta$ .

## Separable, Non-Separable, Dense, and Non-Dense Sets

(1)  $(\mathbb{R}^d, \|\cdot\|_p)$  is separable for any  $p \in [1, \infty]$ . Indeed,  $\mathbb{Q}^d \subseteq \mathbb{R}^d$  is the countable dense subset of  $\mathbb{R}^d$ .

Let 
$$r = \begin{bmatrix} r_1 \\ \vdots \\ r_d \end{bmatrix} \in \mathbb{R}^d$$
. Find  $q = \begin{bmatrix} q_1 \\ \vdots \\ q_d \end{bmatrix} \in \mathbb{Q}^d$  with  $|r_j - q_j| < \varepsilon/d$ . Then,

$$||r - q||_1 = \sum_{j=1}^{d} |r_j - q_j|$$
 $< d.$ 

We know that for any vector  $r \in \mathbb{R}^d$ , we can find a vector q such that

$$||q-r||_p \le c ||q-r||_1$$
,

so for arbitrary p, find q such that  $\|q - r\|_1 < \varepsilon/c$ .

(2) Similarly,  $\mathbb{C}_{\mathbb{Q}} = \{a + bi \mid a, b \in \mathbb{Q}\}$  is also countable, meaning  $\mathbb{C}^d_{\mathbb{Q}} \subseteq \mathbb{C}^d$  is dense and  $\mathbb{C}^d$  is dense.