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## Chapter 1

### Multivariable Calculus

#### 1.1 Linear Algebra

#### 1.1.1 Vector Spaces

**Definition (Normed vector space):** Let V be a vector space. A **norm** is a real valued function  $\|\cdot\|:V\to\mathbb{R}$  which has the following properties

- 1.  $\forall x \in V, ||x|| > 0.$
- 2.  $||x|| = 0 \implies x = 0$ .
- 3.  $\forall x \in V \ \forall \alpha \in \mathbb{F}, \ \|\alpha x\| = |\alpha| \ \|x\|.$
- 4.  $\forall x, y \in V \|x + y\| \le \|x\| + \|y\|$ .

Each normed vector space induces a metric space (V, d) where d(x, y) = ||y - x||.

**Theorem 1.1.** In every normed space  $(V, \|\cdot\|)$  we have

$$|||v|| - ||w||| \le ||v - w||$$

Hence the norm is Lipschitz continuous.

**Definition:** Assume V is a vector space and let  $\|\cdot\|_1$ ,  $\|\cdot\|_2$  be two norms for V. They are said to be equivalent when

$$\exists c_1, c_2 > 0 \ \forall x: \qquad c_1 \|x\|_1 \le \|x\|_2 \le c_2 \|x\|$$

To check if the above definition is indeed an equivalence relation, we must show that following:

Reflexive  $\|\cdot\|_1 \sim \|\cdot\|_1$ .

 $\mathbf{Symmetric} \ \left\| \ \cdot \ \right\|_1 \sim \left\| \ \cdot \ \right\|_2 \implies \left\| \ \cdot \ \right\|_2 \sim \left\| \ \cdot \ \right\|_1.$ 

$$\textbf{Transitive} \ \|\cdot\|_1 \sim \|\cdot\|_2 \,, \ \|\cdot\|_2 \sim \|\cdot\|_3 \implies \|\cdot\|_1 \sim \|\cdot\|_3.$$

**Remark 1.** Equivalent norms induce equivalent metrics, hence they induce the same topology.

**Theorem 1.2.** All norms defined on a finite dimensional vector space V are equivalent.

*Proof.* Let  $\|\cdot\|$  be an arbitrary norm on V and  $\{e_1, e_2, \ldots, e_n\}$  be a basis of V. Let  $\|\cdot\|_2$  be  $L_2$ -norm (Euclidean norm). It will suffice to show  $\|\cdot\| \sim \|\cdot\|_2$ . Let

$$M = \max(\|e_1\|, \dots, \|e_n\|)$$

Take  $x \in V$ , writing  $x = \sum_{i=1}^{n} \xi_i e_i$  we have:

$$||x|| = \left\| \sum_{i=i}^{n} \xi_{i} e_{i} \right\| \leq \sum_{i=1}^{n} |\xi_{i}| ||e_{i}|| \leq M\sqrt{n} ||x||_{2}$$

Taking  $c_2 = M\sqrt{n}$  proves the right inequality. For the left inequality we need the following lemma

**Lemma 1.3.** If V is a normed vector space with  $\|\cdot\|_2$ , as defined above, is viewed as metric space  $(V, \|\cdot\|_2)$  then  $\|\cdot\|: V \to \mathbb{R}$  is continuous.

*Proof.* Let  $x_0 \in V$  and M be defined as above. For any  $\epsilon > 0$  consider  $\delta = \frac{\epsilon}{M\sqrt{n}}$  then if  $\|x - x_0\|_2 < \delta$ 

$$|||x|| - ||x_0||| \le ||x - x_0|| \le M\sqrt{n} \, ||x - x_0|| \le \epsilon$$

Now consider the sphere of radius r=1 centered at 0,  $S_1(0)=S_1=\{x\in V:\|x\|_2=1\}$ . One can show that S is compact. Therefore,  $\|x\|$  assumes its minimum on S. Let  $a=\|x_0\|$  be the minimum. Since  $0\notin S$  then a>0. By letting  $y=x/\|x\|_2$ , we have  $y\in S$  and thus  $a\leq \|y\|$  which is

$$a \|x\|_2 \le \|x\|$$

Taking  $c_1 = a$  proves the theorem.

**Theorem 1.4.** Let  $(V, \|\cdot\|)$  be a normed space. The following are equivalent

- 1. V is finite dimensional.
- 2. every bounded closed set in V is compact.
- 3. the closed unit ball in V is compact.

Proof.

**Example 1.1.** The closed unit ball in the infinite dimensional vector spave  $C([0,1], \mathbb{R})$  with  $||f|| = \max f(x)$  is not compact. Take  $f_n(x) = x^n$ . Obviously  $||f_n|| = 1$ , however  $f_n$  doesn't uniformly converge and hence  $f_n$  doesn't have a limit in  $C([0,1], \mathbb{R})$  with the max norm. Consider the following norm

$$||f||_I = \int_0^1 |f(x)| \, \mathrm{d} x$$

Note that  $\|\cdot\|_I$  and  $\|\cdot\|_{\max}$  are not equivalent. Let g(x)=0 for all  $x\in[0,1]$ . Then

$$||f_n - g||_I = \frac{1}{n+1} \to 0 \quad \text{as } n \to \infty.$$

**Definition (Banach space):** A normed vector space V that is complete is a **Banach space**. A **Hilbert Space** is a Banach space whose norm is induced by an inner product.

Corollary 1.5. Any finite dimensional normed vector space V is a Banach space.

#### 1.1.2 Linear Maps

Let V and W be a vector spaces over  $\mathbb{F}$ . A map  $T:V\to W$  is **linear** if

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

for all  $x, y \in V$  and  $\lambda \in \mathbb{F}$ .

**Definition:** Let  $(V, \|\cdot\|_V)$  and  $(W, \|\cdot\|_W)$  be normed spaces then, a linear transformation  $T: V \to W$  is **bounded** if there exists a constant C > 0 such that

$$||Tv||_W \le C \, ||v||_V$$

for all  $v \in V$ . We denote the set of all linear map from  $V \to W$  as  $\mathcal{L}(V, W)$ . If  $T \in \mathcal{L}(V, W)$  is bijective such that  $T^{-1} \in \mathcal{L}(V, W)$ , then T is called an **isomorphism** and V, W are **isomorphic**. An operator  $T \in \mathcal{L}(V, W)$  is called **isometric** if  $||Tv||_W = ||v||_V$  for all  $v \in V$ .

**Definition:** If  $(V, \|\cdot\|_V)$ ,  $(W, \|\cdot\|_W)$  are normed spaces then the **operator norm** of a linear transformation  $T: V \to W$  is

$$||T|| = \sup \left\{ \frac{||Tv||_W}{||v||_V} \middle| v \neq 0 \right\}$$

**Proposition 1.6.** Let  $T: U \to V$  and  $T': V \to W$  be two linear transformations.

$$||T' \circ T|| \le ||T|| \, ||T'||$$

*Proof.* for an arbitrary non-zero  $x \in U$ 

$$||T' \circ T(x)||_W \le ||T'|| ||Tx||_V \le ||T'|| ||T|| ||x||_U$$

which implies

$$||T' \circ T|| \le ||T|| \, ||T'||$$

**Theorem 1.7.** Let  $(V, \|\cdot\|_V)$  and  $(W, \|\cdot\|_W)$  be normed spaces and  $T: V \to W$  be a linear transformation. The following are equivalent

- 1. ||T|| is finite.
- 2. T is bounded.
- 3. T is Lipschitz continuous.
- 4. T is continuous at a point.
- 5.  $\sup_{\|v\|_V=1} \|Tv\|_W < \infty$ .

*Proof.* item  $1 \Rightarrow$  item 2: Obviously

$$\begin{split} &\frac{\left\|Tv\right\|_{W}}{\left\|v\right\|_{V}} \leq \left\|T\right\| \\ \Longrightarrow &\left\|Tv\right\|_{W} \leq \left\|T\right\| \left\|v\right\|_{V} \end{split}$$

note that if v=0 then Tv=0 as well and thus the last inequality holds for all  $v\in V$ .

item  $2 \Rightarrow$  item 3:

$$||Tv - Tu||_W = ||T(u - v)||_W \le C ||u - v||_W$$

item  $3 \Rightarrow \text{item } 4$ : Trivial.

item  $4 \Rightarrow$  item 5: Let T be continuous at  $u \in V$ . Then there is a  $\delta > 0$  such that

$$\|v-u\|<\delta \implies \|Tv-Tu\|_W=\|T(v-u)\|_W<1$$

Now for an arbitrary non-zero v we have

$$\left\| \left( \frac{\delta v}{2 \left\| v \right\|_{V}} + u \right) - u \right\|_{V} < \delta$$

Therefore

$$\left\| T \left( \frac{\delta v}{2 \left\| v \right\|_{V}} \right) \right\|_{W} < 1$$

$$\left\| T \left( \frac{v}{\left\| v \right\|_{V}} \right) \right\|_{W} < \frac{2}{\delta}$$

item  $5 \Rightarrow$  item 1: Let  $v \in V$  be an arbitrary vector. Then

$$\sup \left\| T \left( \frac{v}{\|v\|_V} \right) \right\|_W < \infty$$

$$\implies \sup \|Tv\|_W < \infty$$

**Theorem 1.8.** If V is a finite dimensional normed vector space then any linear transformation  $T: V \to W$  is continuous.

*Proof.* Since V is finite dimensional, according to Theorem 1.2, any two norms are equivalent. Hence, take  $\|\cdot\|_2$  to be Euclidean norm over a basis  $\{e_1,\ldots,e_n\}$ . Let x be such that  $\|x\|_2 < \delta$  for some  $\delta > 0$ . Therefore,  $|\xi_i| < \delta^2$ 

$$||Tx||_W = \left\| \sum \xi_i T(e_i) \right\|_w \le \sum |\xi_i| ||T(e_i)||_W \le \delta^2 K$$

where  $K = \max ||T(e_i)||_W$ . By letting  $\delta = \sqrt{\frac{\epsilon}{K}}$  we proved continuity at 0 and hence the continuity by Theorem 1.7.

**Theorem 1.9.** For two normed vector spaces  $V, W, (\mathcal{L}(V, W), ||T||)$  is a normed vector space. Moreover, it is a Banach space when W is a Banach space.

Proof.

**Theorem 1.10.** Let  $(V, \|\cdot\|)$  be a normed space. Then any linear transformation  $T : \mathbb{R}^n \to V$  is continuous. Furthermore, if T is a bijection, it is a heomeomorphis.

Proof.

**Theorem 1.11.**  $f: \mathbb{R}^n \to \mathbb{R}^n$  linear transformation is invertible if and only if there exists a c such that:

$$c \|x\| \le \|f(x)\|$$

*Proof.* A linear transformation  $f: \mathbb{R}^n \to \mathbb{R}^n$  is one-to-one if and only if it is surjective because dim Im f + dim ker f = n. Hence, we only need to show that f is one-to-one.

#### Exercises

1. Show that for a linear transformation T,  $||T|| = \sup_{\|v\|_{V} \le 1} ||Tv||_{W}$ .

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#### 1.2 Derivative

Let V, W be finite dimensional vector spaces and  $f: U \subset V \to W$  where U is open. Then f is differentiable at  $x_0$  when a linear transformation  $T: V \to W$  such that

$$\lim_{\|h\| \to 0} \frac{\|f(x_0 + h) - f(x_0) - T(h)\|}{\|h\|} = 0$$

Or equivalently there exists a sublinear function R(h) such that

$$f(x_0 + h) - f(x_0) - Th = R(h)$$
  $\frac{R(h)}{\|h\|} \to 0$ 

T if it exists is unique, represented by  $f'(x_0)$ , D f, or d f(x) and called the **total derivative** or **Fréchet derivative**.

**Proposition 1.12.** Differentiability of f at x implies continuity at x.

**Proposition 1.13.** Assume  $f:U\subset V\to W$  is differentiable at  $x_0$  and let  $u\in V$  be a non-zero vector then

$$\lim_{t \to 0} \frac{f(x_0 + tu) - f(x_0)}{t} = f'(x_0)(u)$$

*Proof.* Let h = tu then

$$R(tu) = f(x_0 + tu) - f(x_0) - T(tu)$$

$$= f(x_0 + tu) - f(x_0) - tT(u)$$

$$\implies \frac{R(tu)}{t} = \frac{f(x_0 + tu) - f(x_0)}{t} - T(u)$$

$$\implies \lim_{t \to 0} \frac{f(x_0 + tu) - f(x_0)}{t} = T(u)$$

**Definition (Directional derivative):** If we let ||u|| = 1 then the limit in Propostion 1.13 becomes the **directional derivative** of f in the direction of u and is denoted by  $D_u f$ .

**Remark 2.** The existence of all directional derivatives of f doesn't imply its differentiability or even its continuity.

**Theorem 1.14.**  $f: U \subset \mathbb{R}^n \to \mathbb{R}^m$  has all of its partial derivative and they're continuous then f is differentiable.

**Proposition 1.15.** Let  $f, g: V \to W$  be differentiable at x and  $h: W \to U$  be differentiable at y = f(x). Furthermore, let c be an scalar then

- 1. D(f + cg) = D(f) + cD(g).
- 2.  $h \circ f$  is differentiable at x and

$$D(h \circ f) = (D(h) \circ f) \circ D(f)$$

**Proposition 1.16.**  $f: U \subset \mathbb{R}^n \to \mathbb{R}^m$  is differentiable at  $x_0$  if and only if every component is differentiable at  $x_0$ .