

# Infinitesimal Calculus 3

Lecture 1, Sunday October 23, 2022  
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Just like how the focus of Calculus 1 and 2 were of single value functions:

$$f: X \subseteq \mathbb{R} \longrightarrow \mathbb{R}$$

The focus of this course will be on functions:

$$f: X \subseteq \mathbb{R} \longrightarrow \mathbb{R}^m$$

## 1.1 A Soft Review of Linear Algebra

### Definition 1.1.1:

We will recall that a **normed vector space**  $V$  is a vector space equipped with a norm function:

$$\|\cdot\| : W \longrightarrow \mathbb{R}$$

Which satisfies the following axioms:

- Positivity:  $\|v\| \geq 0$  for every  $v \in V$ .
- Homogeneity:  $\|\alpha v\| = |\alpha| \cdot \|v\|$  for every  $\alpha \in \mathbb{R}$  and  $v \in V$ .
- The Triangle Inequality:  $\|v + u\| \leq \|v\| + \|u\|$  for every  $v, u \in V$ .

Notice that by the triangle inequality:

$$\|v - u\| + \|u\| \geq \|v\| \quad \|v - u\| + \|v\| \geq \|u\|$$

So we have that

$$\|v - u\| \geq |\|v\| - \|u\||$$

### Example:

Some examples of norms over  $\mathbb{R}^n$  are:

- $\|x\|_\infty = \max |x_1|, \dots, |x_n|$
- $\|x\|_p = \left( \sum_{k=1}^n |x_k|^p \right)^{\frac{1}{p}}$  for  $1 \leq p < \infty$ .

These are actually special cases of the more general  $\ell^p$  norm, which itself can be seen as a special case of the  $L^p(\mathbb{R})$  norm.

### Example:

The set:

$$C([a, b]) := \{f: [a, b] \longrightarrow \mathbb{R} \mid f \text{ is continuous}\}$$

is also a vector space, and we can norm it by the  $\|\cdot\|_{\max}$  norm:

$$\|f\|_{\max} := \max_{x \in [a, b]} f(x)$$

We can also define the  $\|\cdot\|_p$  norm:

$$\|f\|_p := \left( \int_a^b |f(x)|^p dx \right)^{\frac{1}{p}}$$

This too is a specific case of the  $L^p[a, b]$  norm.

Note that  $C[a, b]$  is infinite-dimensional. We can show this by showing that the set  $\{x^n\}_{n \in \mathbb{N}} \subseteq C[a, b]$  is linearly independent. Suppose then that there is a finite sum of  $x^n$ s which equals 0:

$$\sum_{k=1}^n a_k x^{k_n} = 0$$

Notice that the sum above is a polynomial, we will let it equal  $p(x)$ , so  $p \equiv 0$ . But we know that a polynomial is identically 0 if and only if its coefficients are all 0, meaning that  $a_k = 0$ . So  $\{x^n\}_{n \in \mathbb{N}}$  is indeed linearly independent, so  $C[a, b]$  is infinite dimensional.

### Definition 1.1.2:

Again recall that given a vector space  $V$ , an **inner product** on  $V$  is a function:

$$\langle \cdot, \cdot \rangle : V \times V \longrightarrow \mathbb{R}$$

Such that:

- $\langle v, v \rangle \geq 0$  and is 0 if and only if  $v = 0$ .
- $\langle \alpha v, u \rangle = \alpha \langle v, u \rangle$
- $\langle v, u \rangle = \overline{\langle u, v \rangle}$ . Since we are working over  $\mathbb{R}$  we can simplify this to  $\langle v, u \rangle = \langle u, v \rangle$ .

If  $V$  has an inner product, it is called an **inner product space**.

An inner product also generates a norm, we can define:

$$\|v\| = \sqrt{\langle v, v \rangle}$$

### Example:

Over  $C[a, b]$  we can define:

$$\langle f, g \rangle = \int_a^b f(x) \cdot g(x) dx$$

This inner product actually generates the  $\|\cdot\|_2$  norm.

### Example:

The  $\ell^p$  space is the space of all infinite sequences  $\{a_n\}_{n=1}^\infty$  such that:

$$\sum_{n=1}^{\infty} |a_n|^p < \infty$$

And we define the  $\ell^p$  norm, also denoted  $\|\cdot\|_p$  to be the  $p$ th root of this.

Over  $\ell^2$  we can define an inner product:

$$\langle \{a_n\}_{n=1}^\infty, \{b_n\}_{n=1}^\infty \rangle = \sum_{n=1}^{\infty} a_n \cdot b_n$$

It can be seen that this is well-defined and it is trivial to see that this generates the  $\ell^2$  norm.

### Definition 1.1.3:

Given a normed vector space  $V$ , we can define the **distance metric** to be:

$$d(v, u) = \|v - u\|$$

**Definition 1.1.4:**

A set  $M$  equipped with a function:

$$d(\cdot, \cdot): M \times M \longrightarrow \mathbb{R}_{\geq 0}$$

Which satisfies the following:

- Positivity:  $d(v, u) \geq 0$  and is 0 if and only if  $v = u$ .
- Symmetry:  $d(v, u) = d(u, v)$ .
- The Triangle Inequality:  $d(v, u) \leq d(v, w) + d(w, u)$ .

Notice then that a normed vector space is a metric space since:

$$d(v, w) + d(w, u) = \|v - w\| + \|w - u\| \geq \|v - u\| = d(v, u)$$

And the other two requirements are simple to prove.

**Example:**

Not every metric space is a vector space. We can define the  $\mathbb{S}^n$  space as:

$$\mathbb{S}^n := \{v \in \mathbb{R}^{n+1} \mid \|v\| = 1\} \subset \mathbb{R}^{n+1}$$

We can define a metric on  $\mathbb{S}^n$  to be the length of the smallest arc between two points. This is obviously not a vector space, but it *is* a metric space.