

Emory University  
**MATH 411 & 2 Real Analysis**  
Learning Notes

Jiuru Lyu

June 18, 2025

**Contents**

<b>1</b>	<b>The Real Line and Euclidean Space</b>	<b>4</b>
1.1	Algebraic Properties of $\mathbb{R}$ (as a Ordered Field)	4
1.2	Construction of $\mathbb{R}$ and Completeness of $\mathbb{R}$	7
1.3	Another Approach: Least Upper Bound	11
1.4	Cauchy Sequence and Cauchy Completeness	13
	Summary I: Completeness on Ordered Field	15
1.5	$\liminf$ and $\limsup$	16
1.6	Euclidean Space $\mathbb{R}^n$ and General Metric Space	18
<b>2</b>	<b>Topology of Euclidean Space</b>	<b>23</b>
2.1	Open Set	23
2.2	Closed Sets	26
	Summary II: Definitions of Point Set Topology	28
2.3	Convergence	29
2.4	Completeness	31
<b>3</b>	<b>Compactness and Connectedness</b>	<b>33</b>
3.1	Compactness	33
3.2	Compactness in $\mathbb{R}^n$	38
3.3	Nested Set Property	39
3.4	Connectedness	40
<b>4</b>	<b>Continuous Mappings</b>	<b>44</b>
4.1	Continuity	44
4.2	Properties of Continuous Mappings	46
4.3	Uniform Continuity (UC)	49

4.4	Differentiability . . . . .	51
4.5	Integration . . . . .	57
<b>5</b>	<b>Uniform Convergence</b>	<b>65</b>
5.1	Definition of Convergence . . . . .	65
5.2	Integration and Differentiation of Series . . . . .	69
5.3	The Space of Continuous Functions . . . . .	70
5.4	The Contraction Mapping Principle (CMP) . . . . .	73
<b>6</b>	<b>Differential Mappings</b>	<b>77</b>
6.1	Definition and Matrix Representation of a Differential . . . . .	77
6.2	Necessary and Sufficient Conditions for Differentiability . . . . .	82
	Summary III: Relations among Properties of $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ . . . . .	88
6.3	Differentiation Rules . . . . .	88
6.3.1	Chain Rule . . . . .	88
6.3.2	Other Differentiation Rules . . . . .	90
6.4	Geometric Interpretation of Gradient . . . . .	90
6.5	Mean Value Theorem (MVT) . . . . .	92
6.6	Taylor's Theorem & Higher Order Differentials . . . . .	93
6.6.1	One Dimensional Case . . . . .	93
	Summary IV: Taylor's Formula & Taylor's Approximation . . . . .	94
6.6.2	Taylor Series . . . . .	95
6.6.3	Higher Dimensional Case . . . . .	97
	Summary V: Higher Order Differentials . . . . .	99
6.7	Minima & Maxima in $\mathbb{R}^n$ . . . . .	101
<b>7</b>	<b>Inverse and Implicit Function Theorem</b>	<b>107</b>
7.1	Inverse Function Theorem . . . . .	107
7.2	Implicit Function Thm and Applications . . . . .	112
7.3	Constrained Extrema . . . . .	115
7.3.1	Morse Theory: Local Behavior Near a Critical Point . . . . .	115
7.3.2	Constrained Extremal Problem . . . . .	116
<b>8</b>	<b>Integration</b>	<b>119</b>
8.1	Definition of Integration . . . . .	119
8.2	Criterion for Integrability . . . . .	124
8.2.1	Measure Zero: How to Measure the Size of a Set . . . . .	124
8.2.2	Lebesgue's Theorem . . . . .	126
8.3	Improper Integrals . . . . .	132
8.4	Lebesgue Convergence Theorem . . . . .	136

<b>9 Computing Integrals</b>	<b>140</b>
9.1 Fubini's Theorem . . . . .	140
9.2 Change of Variable . . . . .	143
<b>10 Fourier Analysis</b>	<b>149</b>
10.1 Introduction . . . . .	149
10.2 Inner Product Space of Functions . . . . .	149
10.2.1 Basic Concepts . . . . .	149
10.2.2 The Space $\mathcal{C}$ and $L^2$ . . . . .	151
10.3 Fourier Analysis on Inner Product Space . . . . .	152
10.3.1 Geometry of an Inner Product Space . . . . .	152
10.3.2 Fourier Series and Complete Family . . . . .	153
10.4 Completeness and Convergence in $L^2$ . . . . .	156

# 1 The Real Line and Euclidean Space

## 1.1 Algebraic Properties of $\mathbb{R}$ (as a Ordered Field)

**Axiom 1.1.1 Field Axioms:** Recall the following properties

- Addition Axioms

- (1) *commutativity*:  $x + y = y + x$
- (2) *associativity*:  $(x + y) + z = x + (y + z)$
- (3) *the zero element*:  $x + 0 = x$
- (4) *the negative element*:  $x + (-x) = 0$

This further gives the definition of *subtraction*:  $y - x = y + (-x)$ .

- Multiplication Axioms

- (5) *commutativity*:  $xy = yx$
- (6) *associativity*:  $(xy)z = x(yz)$
- (7) *the one element/unit vector*:  $x \cdot 1 = x$
- (8) *inverse*: for each  $x \neq 0$ ,  $\exists x^{-1}$  s.t.  $x \cdot x^{-1} = 1$

This further gives the definition of *division*:  $y/x = y \cdot x^{-1}$  when  $x \neq 0$ .

- (9) *distribution*:  $x(y + z) = xy + xz$

- (10)  $1 \neq 0$

- Order Axioms

- (11) *reflexivity*:  $x \leq x$
- (12) *anti-symmetry*: If  $x \leq y$  and  $y \leq x \implies x = y$ .
- (13) *transitivity*: If  $x \leq y$  and  $y \leq z \implies x \leq z$
- (14) *linear relation*: For each pair  $x, y$ , either  $x \leq y$  or  $y \leq x$ .
- (15) *compatibility with addition*: If  $x \leq y \implies x + z \leq y + z \quad \forall z$
- (16) *compatibility with multiplication*: If  $0 \leq x$  and  $0 \leq y \implies 0 \leq xy$ .

**Definition 1.1.2 (Ordered Field).** A system (or a set)  $\mathcal{F}$  is called an *ordered field* if it satisfies all the above 16 properties.

**Remark 1.1 (Examples of Ordered Field)**  $\mathbb{R}$  and  $\mathbb{Q}$ .

**Definition 1.1.3 (Field).** A set is called a *field* if satisfies all the addition and multiplication axioms.

**Definition 1.1.4 (Ring).** A set is a *ring* if it satisfies (1) – (9) except (5) and (8).

**Example 1.1.5  $\mathbb{Z}$  as a Ring**

$\mathbb{Z} = \{\dots, -3, -2, -1, 0, 1, 2, 3, \dots\}$ , the set of integers, is a commutative ring, but not a field.

**Remark 1.2** *There is no division operation in a ring as multiplicative inverse is not defined.*

**Definition 1.1.6 (Group).** A set is a *group* if it satisfies (1) – (4).

**Theorem 1.1.7 Law of Trichotomy**

If  $x$  and  $y$  are elements of an ordered field, then exactly one of the relations  $x < y$ ,  $x = y$ , or  $x > y$  holds.

**Proposition 1.1.8 Other Algebraic Properties of  $\mathbb{R}$  (as an Ordered Field):**

1. *unique identities*: If  $a + x = a$  for every  $a$ , then  $x = 0$ . If  $a \cdot x = a$  for every  $a$ , then  $x = 1$ .
2. *unique inverses*: If  $a + x = 0$ , then  $x = -a$ . If  $ax = 1$ , then  $x = a^{-1}$ .
3. *no divisors of zero*: If  $xy = 0$ , then  $x = 0$  or  $y = 0$ .
4. *cancellation laws for addition*: If  $a + x = b + x$ , then  $a = b$ . If  $a + x \leq b + x$ , then  $a \leq b$ .
5. *cancellation for multiplication*: If  $ax = bx$  and  $x \neq 0$ , then  $a = b$ . If  $ax \geq bx$  and  $x > 0$ , then  $a \geq b$ .
6.  $0 \cdot x = 0$  for every  $x$ .
7.  $-(-x) = x$  for every  $x$ .
8.  $-x = (-1)x$  for every  $x$ .
9. If  $x \neq 0$ , then  $x^{-1} \neq 0$  and  $(x^{-1})^{-1} = x$ .
10. If  $x \neq 0$  and  $y \neq 0$ , then  $xy \neq 0$  and  $(xy)^{-1} = x^{-1}y^{-1}$ .
11. If  $x \leq y$  and  $0 \leq z$ , then  $xz \leq yz$ . If  $x \leq y$  and  $z \leq 0$ , then  $yz \leq xz$ .
12. If  $x \leq 0$  and  $y \leq 0$ , then  $xy \geq 0$ . If  $x \leq 0$  and  $y \geq 0$ , then  $xy \leq 0$ .
13.  $0 < 1$ .
14. For any  $x$ ,  $x^2 \geq 0$ .

**Proof 1.** (Of No. 14)

**Case I** If  $x \geq 0$ , then  $x^2 = x \cdot x \geq 0$ , by property (16) of Axiom 1.1.

**Case II** If  $x < 0$ , then

$$\begin{aligned} x^2 &= x \cdot x = (-1)(-x) \cdot (-1)(-x) \quad [\text{by property 7 of Proposition 1.7}] \\ &= (-1)^2 \cdot (-x)^2. \end{aligned}$$

Note that  $0 = (-1)(-1 + 1) = (-1)^2 + (-1)$  if we distribute  $(-1)$ . Then, adding 1 on both sides, we have

$$1 = (-1)^2 + (-1) + 1 = (-1)^2 \quad [\text{by additive inverse}]$$

That is,  $(-1)^2 = 1$ . So,  $x^2 = (-1)^2 \cdot (-x)^2 = 1 \cdot (-x)^2 = (-x)^2 \geq 0$  by Case I.

Q.E.D. ■

**Proposition 1.1.9:**  $ab \leq \frac{a^2 + b^2}{2}$ .

**Proof 2.**

$$(a - b)^2 \geq 0 \quad [\text{By property 14 of Proposition 1.7}]$$

$$a^2 + b^2 - 2ab \geq 0$$

$$2ab \leq a^2 + b^2$$

$$ab \leq \frac{a^2 + b^2}{2}.$$

Q.E.D. ■

**Definition 1.1.10 (Absolute Value (Norm) and Distance (Metric)).** For  $x, y \in \mathbb{R}$ ,  $|x| = \begin{cases} x, & x \geq 0 \\ -x, & x < 0 \end{cases}$

is the *absolute value*, and  $d(x, y) = |x - y|$  is the *distance*.

**Proposition 1.1.11 Properties of Absolute Value and Distance:**

- $|x| \geq 0$  for every  $x$ .
- $|x| = 0$  if and only if  $x = 0$ .
- $|xy| = |x||y|$ .
- $d(x, y) \geq 0$
- $d(x, y) = 0$  if and only if  $x = y$ .
- $d(x, y) = d(y, x)$ .

### Theorem 1.1.12 Triangle Inequalities

$\forall x, y, z \in \mathbb{R}$

1.  $|x + y| \leq |x| + |y|$
2.  $||x| - |y|| \leq |x - y|$
3.  $d(x, y) \leq d(x, z) + d(z, y)$

**Proof 3.** (Of No. 1)

**Case I** Suppose  $x \geq 0$  and  $y \geq 0$ . Then,  $x + y \geq 0$ , and

$$|x + y| = x + y = |x| + |y|. \quad \square$$

**Case II** WLOG, suppose  $x \geq 0$  and  $y < 0$ .

- Suppose  $x + y \geq 0$ , then

$$|x + y| = x + y = |x| - (-y) = |x| - |y| \leq |x| + |y|. \quad \square$$

- Suppose  $x + y < 0$ , then

$$|x + y| = -(x + y) = -x - y = -|x| + |y| \leq |x| + |y|. \quad \square$$

**Case III** Suppose  $x < 0$  and  $y < 0$ . Then,  $x + y < 0$ , and

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

Q.E.D. ■

## 1.2 Construction of $\mathbb{R}$ and Completeness of $\mathbb{R}$

**Notation 1.1.** Recall the following number systems:

$$\begin{array}{l|l} \mathbb{N} = \mathbb{Z}^+ = \{0, 1, 2, 3, \dots\} & \text{non-negative integers} \\ \mathbb{Z} & \text{integers} \\ \mathbb{Q} = \left\{ \frac{m}{n} : m, n \in \mathbb{Z}, n \neq 0 \right\} & \text{rational numbers} \end{array}$$

**Proposition 1.2.2 Important Properties of Number Systems:**

- For  $\mathbb{N}$ :
  - **Definition 1.2.3 (Principle of mathematical induction).** If  $S$  is a subset of  $\mathbb{Z}^+$  s.t.  $0 \in S$  and  $k \in S \implies k + 1 \in S$ , then  $S = \mathbb{Z}^+$ .
  - **Definition 1.2.4 (Well-Ordered Property).** Each subset  $S \neq \emptyset$  has a smallest element.  
As a consequence of well-ordering property, we have the principle of complete induction:
  - **Definition 1.2.5 (Principle of Complete Induction).** If  $S \subset \mathbb{Z}^+$  is a subset s.t.  $\{x \in \mathbb{Z}^+ \mid x < n\} \subset S \implies n \in S$ , then  $S = \mathbb{Z}^+$ .
- For  $\mathbb{Z}$ :
  - Commutative ring with identity
- For  $\mathbb{Q}$ :
  - **Definition 1.2.6 (Countable).**  $\mathbb{Q}$  can be placed in one-to-one correspondence with  $\mathbb{N}$  (or a subset of it). The whole  $\mathbb{Q}$  can be displayed as a list or sequence.

**Remark 1.3** A simple way to prove it is to consider the points in the plane with integer coordinates, say  $(p, q)$ . After assigning fraction  $\frac{p}{q}$  (simplified to lowest terms and leave out cases when  $q = 0$ ) to this point, we achieve a one-to-one correspondence.

- **Definition 1.2.7 (Dense in Itself).** If  $x, y \in \mathbb{Q}$  and  $x < y \implies \exists z \in \mathbb{Q} \text{ s.t. } x < z < y$ .
- **Proposition 1.2.8 Archimedean Property:**

$$\forall x \in \mathbb{Q}, \exists n \in \mathbb{Z} \text{ s.t. } n > x.$$

**Proof 1.** If  $x \leq 0$ , take  $n = 1$ . If  $x = \frac{p}{q}$  with  $p, q > 0$ , take  $n = p + 1$ .

Q.E.D. ■

**Remark 1.4** *Equivalent formulation of the Archimedean Property:*

- \* If  $x \in \mathbb{Q}$ , then  $\exists$  integer  $n$  s.t.  $x < n$ .
- \* If  $x, y \in \mathbb{Q}$  and  $0 < x < y$ , then  $\exists$  integer  $k$  s.t.  $kx > y$ .
- \* If  $x > 0 \in \mathbb{Q}$ , then  $\exists$  integer  $n > 0$  s.t.  $0 < \frac{1}{n} < x$ .

- Ordered field.

**$\mathbb{Q}$  is already an ordered field, why do we bother to define  $\mathbb{R}$  for analysis?**

The big idea:  $\mathbb{Q}$  is not *quite complete*

- Evidence 1 (Analysis POV): There is no rational whose square is 2. That is,  $x^2 = 2$  has no solution in  $\mathbb{R}$ .

**Proof 2.** We will use proof by contradiction. Assume  $\exists$  solution  $x = \frac{m}{n}$  with  $m, n \in \mathbb{Z}$  and they have no common factors. Then,

$$\left(\frac{m}{n}\right)^2 = 2 \implies m^2 = 2n^2.$$

So,  $m^2$  is even, then  $m$  is even as well. Suppose  $m = 2k$ ,  $k \in \mathbb{Z}$ . Then,

$$\begin{aligned} m^2 &= (2k)^2 = 4k^2 = 2n^2 \\ n^2 &= 2k^2. \end{aligned}$$

So,  $n^2$  is even, and  $n$  is even.

\*  $m, n$  both even, so they have a common factor of 2. This contradict with our assumption.

So,  $\nexists$  a solution  $x \in \mathbb{Q} \text{ s.t. } x^2 = 2$ .

Q.E.D. ■

- Evidence 2 (Geometry POV): There is no rational representation of the diagonal of a square of size 1.

**Remark 1.5 (Informal Definition of Sequence Limit)** *A sequence is said to converge to a limit  $x$  if we can guarantee that the points in the sequence are as close as we wish to  $x$  by going far enough out in the sequence.*



**Definition 1.2.9 (Limit of a Sequence).** A sequence  $\{x_n\}$  is said to *converge* to  $x$  if  $\forall \varepsilon > 0, \exists$  integer  $N$  s.t.  $|x_n - x| < \varepsilon$  whenever  $n \geq N$ . (Alternatively,  $n \geq N \implies |x_n - x| < \varepsilon$ ). We denote the *limit* as

$$\lim_{n \rightarrow \infty} x_n = x \quad \text{or, simply} \quad x_n \rightarrow x \text{ as } n \rightarrow \infty.$$

**Remark 1.6**  $N$  depends on  $\varepsilon$ , and the smaller the  $\varepsilon$ , the bigger the  $N$ .

**Example 1.2.10 Show**  $\lim_{n \rightarrow \infty} \frac{n+1}{n+2} = 1$ .

**Proof 3.** Given  $\varepsilon > 0$  [fix  $\varepsilon$ ], we need to find  $N$  s.t.  $n \geq N \implies |x_n - 1| < \varepsilon$ , where  $x_n = \frac{n+1}{n+2}$ .

Consider

$$|x_n - 1| = \left| \frac{n+1}{n+2} - 1 \right| = \left| \frac{n+1 - n - 2}{n+2} \right| = \left| \frac{-1}{n+2} \right| = \frac{1}{n+2}.$$

Then, we want

$$\frac{1}{n+2} < \varepsilon \iff n+2 > \frac{1}{\varepsilon} \iff n > \frac{1}{\varepsilon} - 2.$$

By the Archimedean property, choose integer  $N > \frac{1}{\varepsilon} - 2$ . [ $N$  is fixed and is what we want to find]

Then, based on the arguments, when  $n \geq N$  [ $n$  is changing], we have

$$|x_n - 1| = \frac{1}{n+2} \leq \frac{1}{N+2} < \varepsilon.$$

That is,

$$\lim_{n \rightarrow \infty} x_n = 1.$$

Q.E.D. ■

**Theorem 1.2.11 Basic Properties of Limits**

- **Sandwich Lemma/Squeeze Theorem:** Suppose  $x_n \rightarrow L$ ,  $y_n \rightarrow L$ , and  $x_n \leq z_n \leq y_n$  for all  $n$ . Then,  $z_n \rightarrow L$ . It is also enough to assume that  $\exists N_0$  s.t.  $n > N_0 \implies x_n \leq z_n \leq y_n$
- If  $a \leq x_n \leq b$  for every  $n$  and  $x_n \rightarrow x$ , then  $a \leq x \leq b$ .
- **Uniqueness:** If  $x_n$  is a sequence in an ordered field and  $x_n \rightarrow x$  and  $x_n \rightarrow y$ , then  $x = y$ .
- **Boundedness:** A convergent sequence is bounded.
- **Arithmetic of Sequence and Limits:** Suppose  $x_n \rightarrow x$  and  $y_n \rightarrow y$ . Then,

$$\{x_n\} + \{y_n\} = \{x_n + y_n\} \implies x_n + y_n \rightarrow x + y$$

$$\lambda\{x_n\} = \{\lambda x_n\} \implies \lambda x_n \rightarrow \lambda x$$

$$\{x_n\}\{y_n\} = \{x_n y_n\} \implies x_n y_n \rightarrow xy$$

$$\{x_n\}/\{y_n\} = \{x_n/y_n\} \implies x_n/y_n \rightarrow x/y$$

**Definition 1.2.12 (Monotone Sequence Property/MSP).** Every *monotone increasing sequence* that is *bounded (bdd) above* converges.

**Remark 1.7** “*monotone increasing sequence*” refers to a sequence where  $x_n \leq x_{n+1} \quad \forall n$ ; “*bdd above*” refers to  $\exists x$  s.t.  $x_n \leq x \quad \forall n$ , and we call this  $x$  an upper bound.

**Definition 1.2.13 (Completeness).** An ordered field  $\mathcal{F}$  is said to be *complete* if it has the MSP.

**Construction of  $\mathbb{R}$  (from  $\mathbb{Q}$ )**

Consider set  $S$  of sequences,

$$S = \{(x_1, x_2, \dots) \mid x_n \in \mathbb{Q}, x_n \uparrow \text{ (monotone increasing)}, x_n \text{ bdd above}\}.$$

Define *equivalence relation* (reflexive, transitive, symmetric)  $\sim$  on  $S$ :

$$\{x_n\} \sim \{y_n\} \iff x_n \text{ and } y_n \text{ have the same upper bounds.}$$

Then, each equivalence class defines a unique real number (as the limit of the representing sequence). Let

$$\mathbb{R} = \{x \mid x \text{ is an equivalence class in } S\}.$$

If  $r \in \mathbb{Q}$ , then  $r$  is represented by the sequence  $r$  itself ( $\{r\}$ ). So,  $\mathbb{Q} \subseteq \mathbb{R}$ .

**Claim 1.2.14**  $\mathbb{R}$  is a complete ordered field under the following operations: For  $x = [\{x_n\}]$  and  $y = [\{y_n\}]$ ,

- **Addition:**  $x + y = [\{x_n + y_n\}]$

- Multiplication:  $x \cdot y = [\{x_n \cdot y_n\}]$
- Order:  $x \leq y \iff \exists$  upper bd of  $\{x_n\}$  that is  $\leq$  all upper bd of  $\{y_n\}$ .

**Theorem 1.2.15**

$\mathbb{R}$  is the “unique” complete ordered field.

**Remark 1.8** By unique, we mean isomorphism. That is, if  $\exists$  another complete ordered field  $\mathcal{F}$ , we can put  $\mathcal{F}$  and  $\mathbb{R}$  into a one-to-one relationship.

**Proposition 1.2.16 Properties of  $\mathbb{R}$ :**

- $\mathbb{R}$  is Archimedean:  $\forall x \in \mathbb{R}, \exists$  integer  $n > x$ .
- $\mathbb{Q}$  is dense in  $\mathbb{R}$ :
  - If  $x, y \in \mathbb{R}$  and  $x < y \implies \exists r \in \mathbb{Q}$  s.t.  $x < r < y$ .
  - If  $x \in \mathbb{R}$  and  $\varepsilon > 0 \implies \exists r \in \mathbb{Q}$  s.t.  $|x - r| < \varepsilon$ .
- The interval  $(0, 1)$  is uncountable. (Hence,  $\mathbb{R}$  is uncountable).

**Proof 4.** (of uncountability)

Assume  $(0, 1)$  is countable. Then, it can be put into a one-to-one relationship with  $\mathbb{N}$ . Say the following list exhausts elements of  $\mathbb{R}$ :

$$x_1 = 0.a_{11}a_{12} \cdots a_{1n} \cdots, x_2 = 0.a_{21}a_{22} \cdots a_{2n} \cdots, \dots, x_k = 0.a_{k1}a_{k2} \cdots a_{kn} \cdots, \dots$$

[Goal: find a new number that is not in the list] Define a new number:

$$x = 0.x'_1x'_2 \cdots x'_k \cdots,$$

where for each  $k$ ,  $x'_k = \begin{cases} 4 & \text{if } a_{kk} \neq 4 \\ 3 & \text{if } a_{kk} = 4. \end{cases}$  [This construction ensures  $x'_k \neq a_{kk}$ ] Then,  $x \in (0, 1)$  and

$x \neq x_k \quad \forall k$ . \* We have constructed a number that is not in the list. So,  $(0, 1)$  is not countable.

Q.E.D. ■

**1.3 Another Approach: Least Upper Bound**

**Definition 1.3.1 (Upper Bound/Least Upper Bound).** Let  $S \subset \mathbb{R}$ .

- We say  $b$  is an *upper bd* for  $S$  if  $x \leq b \quad \forall x \in S$ .
- We say  $b$  is a *least upper bd* for  $S$  if  $b$  is an upper bd and  $\leq$  any upper bd of  $S$ .

We use  $\text{lub}(S) = \sup(S)$  to denote the least upper bd. ( $\sup$  stands for supremum). For sets without an upper bound, we define  $\sup(S) = +\infty$ .

**Remark 1.9**  $b = \text{lub}(S) \iff (1) b \text{ is an upper bound, and } (2) b \leq \text{any upper bound of } S.$

**Example 1.3.2**

Suppose  $S_1 = (0, 2)$ ;  $S_2 = [0, 2]$ ;  $S_3 = \emptyset$ ;  $S_4 = (0, \infty)$ . Then,  $\text{lub}(S_1) = 2$ ,  $\text{lub}(S_2) = 2$ ,  $\text{lub}(S_3) = +\infty$ ,  $\text{lub}(S_4) = +\infty$ .

**Definition 1.3.3 (Greatest Lower Bound).** We use  $\text{glb}(S) = \inf(S)$  to denote the greatest lower bound. It is the largest lower bound of  $S$ . For sets without a lower bound, we define  $\inf(S) = -\infty$ .

**Example 1.3.4**

Example 1.3.2:  $\inf(S_1) = 0$ ,  $\inf(S_2) = 0$ ,  $\inf(S_3) = -\infty$ ,  $\inf(S_4) = 0$ ,  $\inf((-\infty, 4)) = -\infty$ .

**Proposition 1.3.5 :** Let  $S \subset \mathbb{R}$ ,  $S \neq \emptyset$ , then

- $b = \text{lub}(S) \iff b \text{ is an upper bound and } \forall \varepsilon > 0, \exists x \in S \text{ s.t. } x > b - \varepsilon. \text{ This implies that an element slightly smaller than } b \text{ is not an upper bound any more.}$
- $a = \inf(S) \iff a \text{ is a lower bound and } \forall \varepsilon > 0, \exists x \in S \text{ s.t. } x < a + \varepsilon.$

**Proposition 1.3.6 :** Suppose  $\emptyset \neq A \subset B \subset \mathbb{R}$ . Then,

$$\inf(B) \leq \inf(A) \leq \sup(A) \leq \sup(B).$$

**Theorem 1.3.7 Equivalent Condition for Completeness: Least Upper Bound Condition**

$\mathbb{R}$  has the following properties:

- LUB property: Every non-empty subset bounded above has the least upper bound.
- GLB property: Every non-empty subset bounded below has the greatest lower bound.

**Proof 1.** (of the LUB Property)

Set-up: Fix any  $S \subset \mathbb{R}$  that is bounded above and  $S \neq \emptyset$ .

[WTS: the existence of  $\text{lub}(S)$   $\iff$  Tool: MSP (but we need to construct monotone sequence first.)

**Step 1** Construction of a Monotone Sequence

Fix an upper bound  $M$  for  $S$ . For each fixed integer  $n \geq 1$ , consider  $a_k = M - \frac{k}{2^n}, k = 1, 2, \dots$ . By the well-ordering property, we can choose an integer  $k_n$  who is the 1<sup>st</sup> integer  $k$  s.t.  $a_k$  is not an upper bound.

Let  $b_n = M - \frac{k_n}{2^n}$ . Then,  $b_n$  is not an upper bound, but  $b_n + \frac{1}{2^n}$  is an upper bound (by construction).

**Step 2** Apply MSP to  $\{b_n\}$

- $b_n$  is monotone increasing:

Note that

$$b_{n+1} - b_n = \left(M - \frac{k_{n+1}}{2^{n+1}}\right) - \left(M - \frac{k_n}{2^n}\right) = \frac{2k_n - k_{n+1}}{2^{n+1}}$$

Suppose, for the sake of contradiction, that  $b_{n+1} - b_n < 0$ . Then,  $b_{n+1} - b_n \leq -\frac{1}{2^{n+1}}$ . That is,

$$b_n \geq b_{n+1} + \frac{1}{2^{n+1}}.$$

\* However, by construction,  $b_n$  is not an upper bound, but  $b_{n+1} + \frac{1}{2^{n+1}}$  is an upper bound. So, there is a contradiction, and thus  $b_{n+1} - b_n > 0$ . This contradiction shows that  $b_n$  is a monotone increasing sequence.

- $b_n$  is bounded above:

Note that  $b_n \leq M$ . So,  $b_n$  is bounded above.

By MSP, suppose  $b_n \rightarrow b$  for some  $b \in \mathbb{R}$ .

**Step 3** Show  $b = \text{lub}(S)$

- $b$  is an upper bound:

Fix  $x \in S$ , we have  $x \leq b_n + \frac{1}{2^n} \quad \forall n$ . When  $x \rightarrow \infty$ ,  $x \leq b + 0$ . So,  $x \leq b$ .

- $b$  is the least upper bound: [WTS:  $\forall \varepsilon > 0, \exists x \in S \text{ s.t. } b - \varepsilon < x$ .]

As  $b$  is the limit, we can always find a  $b_n$  s.t.  $|b_n - b| < \varepsilon$ . That is,  $b - b_n < \varepsilon$ , or  $b_n > b - \varepsilon$ . Hence,  $b$  is the least upper bound.

Q.E.D. ■

## 1.4 Cauchy Sequence and Cauchy Completeness

**Definition 1.4.1 (Cauchy Sequence).** A sequence  $x_n \in \mathbb{R}$  is a *Cauchy Sequence* if  $\forall \varepsilon > 0, \exists N \text{ s.t. } n, m \geq N \implies |x_n - x_m| < \varepsilon$ .

**Proposition 1.4.2 :** Every convergent sequence is Cauchy.

**Proof 1.** Suppose  $x_n \rightarrow x \in \mathbb{R}$ . Given  $\varepsilon < 0$ . Consider

$$\begin{aligned} |x_n - x_m| &= |x_n - x + x - x_m| \\ &\leq |x_n - x| + |x - x_m| \\ &\leq \frac{\varepsilon}{2} + \frac{\varepsilon}{2}. \end{aligned}$$

Q.E.D. ■

### Theorem 1.4.3 Cauchy Completeness

Every Cauchy sequence in  $\mathbb{R}$  converges.

**Remark 1.10 (Strategy of the Proof)**  $\text{Cauchy Sequence} \xrightarrow{\text{Lemma 1.4.4}} \text{Bounded Sequence} \xrightarrow{\text{Theorem 1.4.5}} \exists \text{ convergent subsequence} + \text{Cauchy sequence} \xrightarrow{\text{Lemma 1.4.6}} \text{Sequence converges}.$

**Lemma 1.4.4 :** Every Cauchy sequence is Bounded.

**Theorem 1.4.5**

Every bounded sequence in  $\mathbb{R}$  has a subsequence that converges to some point in  $\mathbb{R}$ .

**Proof2.** Let  $\{x_n\}$  be a bounded sequence in  $\mathbb{R}$ . Fix  $M$  s.t.  $-M < x_n < M \quad \forall n$ .

Divide  $[-M, M]$  into subintervals  $[-M, 0]$  and  $[0, M]$ . One of them, called  $I_0$ , must contain infinitely many terms of  $\{x_n\}$ . Choose  $n_0$  s.t.  $x_{n_0} \in I_0$ .

Divide  $I_0$  into two equal subintervals. One of them, denoted  $I_1$ , contains infinitely many elements. Choose  $n_1 > n_0$  s.t.  $x_{n_1} \in I_1$ .

Continuing this process, we obtain subintervals  $I_k = [a_k, b_k]$  for  $k = 0, 1, \dots$ , and includes  $n_k$  with the following properties:

- $I_0 \supset I_1 \supset I_2 \supset \dots$
- $b_k - a_k = \frac{M}{2^k}$
- $x_{n_k} \in I_k$

[To prove  $\{x_{n_k}\}$  converges, we prove  $\{a_k\}$  and  $\{b_k\}$  converge, and apply the Squeeze Theorem.]

- **Show  $\{a_k\}$  converges:**  $a_k$  is monotone increasing and bounded. By MSP,  $a_k \rightarrow a \in \mathbb{R}$ .
- **Show  $\{b_k\}$  converges:** Note that  $b_k = a_k + \frac{M}{2^k}$ . When  $k \rightarrow \infty$ ,

$$a_k + \frac{M}{2^k} = a + 0 = a.$$

So,  $b_k \rightarrow a$  when  $k \rightarrow \infty$ .

Hence, as  $a_k \leq x_{n_k} \leq b_k$ ,  $a_k \rightarrow a$ ,  $b_k \rightarrow a$ , it must be that  $x_{n_k} \rightarrow a$  as well.

Q.E.D. ■

**Lemma 1.4.6 :** If a subsequence of a Cauchy sequence converges to  $x$ , then the sequence itself converges to  $x$ .

**Proof3.** Given  $\{x_n\}$  is Cauchy and  $x_{n_k} \rightarrow x$ , [WTS:  $x_n \rightarrow x$ ]. Consider

$$\begin{aligned} |x_n - x| &= |x_n - x_{n_k} + x_{n_k} - x| \\ &\leq \underbrace{|x_n - x_{n_k}|}_{\text{Cauchy} \Rightarrow \text{small}} + \underbrace{|x_{n_k} - x|}_{\text{Convergent} \Rightarrow \text{small}} \end{aligned}$$

Q.E.D. ■

### Summary I: Completeness on Ordered Field

Let  $\mathcal{F}$  be an ordered field.

#### Definitions

- **Archimedean Property:**  $\forall x \in \mathcal{F}, \exists \text{ integer } N \text{ s.t. } x < N$ .  
(Equivalently,  $\forall \varepsilon > 0, \exists \text{ integer } n \text{ s.t. } 0 < \frac{1}{n} < \varepsilon$ ).
- **Monotone Sequence Property (MSP):** Every monotone increasing sequence bounded above converges.
- **Completeness:** We say  $\mathcal{F}$  is complete if it has the MSP.
- **LUB Property:** Every set  $S \neq \emptyset$  bounded above has a least upper bound.
- **Cauchy Property:** Every Cauchy sequence converges.

#### Facts in any ordered field

- $\text{MSP} \implies \text{Archimedean Property}$

**Remark 1.11** In general, the converse is not true. For example,  $\mathbb{Q}$  has the Archimedean property but not MSP.

- $\text{MSP} \iff \text{LUB Property}$ .
- $\text{MSP} \implies \text{Cauchy Property}$

**Remark 1.12** The converse is true when Archimedean property is true.

#### Facts in $\mathbb{R}$

- $\text{MSP} \iff \text{LUB Property} \iff \text{Cauchy Property}$

1.5  $\liminf$  **and**  $\limsup$ **Example 1.5.1 Cluster Points of a Sequence**

Consider the sequence

$$a_n = (-1)^n + \frac{1}{n}.$$

Then,  $a_1 = 0$ ,  $a_2 = 1 + \frac{1}{2}$ ,  $a_3 = -1 + \frac{1}{3}$ ,  $a_4 = 1 + \frac{1}{4}$ ,  $\dots$ . This sequence does not converge. However, its terms seem to “cluster” around 1 and  $-1$ .

**Definition 1.5.2 (Cluster Points).** A point  $x$  is called a *cluster point* of a sequence  $\{x_n\}$  if  $\forall \varepsilon > 0$ ,  $\exists$  infinitely many values of  $n$  s.t.  $|x_n - x| < \varepsilon$ .

**Remark 1.13** *This definition is weaker than that of limits.*

**Proposition 1.5.3 Relation Between Limits and Cluster Points:** Suppose  $x_n \in \mathbb{R}$  and  $x \in \mathbb{R}$ . Then,

1.  $x$  is a cluster point of  $\{x_n\} \iff \forall \varepsilon > 0$  and  $\forall$  integer  $N$ ,  $\exists n > N$  s.t.  $|x_n - x| < \varepsilon$ .
2.  $x$  is a cluster point of  $\{x_n\} \iff \exists$  subsequence  $x_{n_k} \rightarrow x$ .
3.  $x_n \rightarrow x \iff$  every subsequence converges to  $x$ .
4.  $x_n \rightarrow x \iff$  the sequence is bounded and  $x$  is the only cluster point.
5.  $x_n \rightarrow x \iff$  every subsequence has a further sequence that converges to  $x$ .

**Proof 1.** (of some claims)

1. Follows from Definition.
2. ( $\Rightarrow$ ) Assume  $x$  is a cluster point. [WTS:  $\exists$  subsequence  $x_{n_k} \rightarrow x$ ].

Given  $\varepsilon_1 = 1$  and  $N = 1$ , by (1),  $\exists n_1 > 1$  s.t.  $|x_{n_1} - x| < \varepsilon = 1$ .

Given  $\varepsilon_2 = \frac{1}{2}$  and  $N = n_1$ , by (1),  $\exists n_2 > n_1$  s.t.  $|x_{n_2} - x| < \varepsilon = \frac{1}{2}$ .

So, in general, given  $\varepsilon_k = \frac{1}{k}$  and  $N = n_{k-1}$ ,

$$\exists n_k > n_{k-1} = N_k \text{ s.t. } |x_{n_k} - x| < \varepsilon_k = \frac{1}{k}.$$

Then,  $x_{n_k} \rightarrow x$  as  $k \rightarrow \infty$ .

3. ( $\Leftarrow$ ) [Prove by contrapositive/contradiction] Assume every subsequence converges. For the sake of contradiction, assume  $x_n$  does not converge to  $x$ . Then we need to construct a subsequence  $x_{n_k}$  s.t.  $x_{n_k} \not\rightarrow x$ .
4. ( $\Leftarrow$ ) [Prove by contrapositive/contradiction]



5. ( $\Leftarrow$ ) Use (4). Every subsequence has its own subsequence that converges to  $x$ . So,  $x$  is a cluster point of every subsequence. Then, we just need to show  $x$  is the only cluster point of  $\{x_n\}$ .

Q.E.D. ■

**Definition 1.5.4** ( $\liminf$  **and**  $\limsup$ ). Given a sequence  $x_n \in \mathbb{R}$ . For each integer  $k \geq 1$ , let

$$a_k = \inf \underbrace{\{x_{k+1}, x_{k+2}, \dots\}}_{\text{Set } S_k} \quad \text{and} \quad b_k = \sup \{x_{k+1}, x_{k+2}, \dots\} = \sup S_k.$$

Then,

$$\liminf x_n = \sup \{a_k\} \quad \text{and} \quad \limsup x_n = \inf \{b_k\}.$$

**Remark 1.14**

- $a_k \leq b_k$ ,  $a_k$  is monotone increasing sequence, and  $b_k$  is monotone decreasing sequence. Thus,

$$\liminf x_n = \lim_{k \rightarrow \infty} a_k \quad \text{and} \quad \limsup x_n = \lim_{k \rightarrow \infty} b_k.$$

Also,  $\liminf x_n \leq \limsup x_n$ .

- $\limsup x_n = +\infty \iff b_k = +\infty \quad \forall k \iff x_n$  is not bounded above.

$\liminf x_n = -\infty \iff a_k = -\infty \quad \forall k \iff x_n$  is not bounded below.

**Proposition 1.5.5 :**  $\limsup x_n = b \in \mathbb{R} \iff \forall \varepsilon > 0$ ,

1.  $\exists N$  s.t.  $n \geq N \implies x_n < b + \varepsilon$ , and
2.  $\forall M, \exists n \geq M$  s.t.  $x_n > b - \varepsilon$ .

**Proof2.** (of forward direction) By definition, we know  $\lim_{k \rightarrow \infty} b_k = b$ , which implies  $\forall \varepsilon > 0, \exists N$  s.t.  $k \geq N \implies |b_k - b| < \varepsilon$ . That is,  $-\varepsilon < b_k - b < \varepsilon$ . As  $b_k$  is monotone decreasing,  $b_k - b \geq 0$ . So,  $0 \leq b_k - b < \varepsilon$ .

1. Note that  $b_k = \sup \{x_{k+1}, x_{k+2}, \dots\}$ . So, if  $n > k$ ,  $x_n \leq b_k < b + \varepsilon \quad \forall k \geq N$ . Therefore,

$$n \geq N + 1 \implies x_n < b + \varepsilon.$$

2. We have  $0 \leq b_k - b$ , or  $b_k \geq b \quad \forall k$ . Given any integer  $M$ . [We need to find  $n \geq M$  s.t.  $x_n > b - \varepsilon$ ]  
Then,

$$b_M = \sup \{x_{M+1}, x_{M+2}, \dots\} \geq b.$$

So, by definition of supremum, we can find  $n > M$  s.t.  $x_n > b_M - \varepsilon \geq b - \varepsilon$ .

Q.E.D. ■

**Proposition 1.5.6 :**  $\limsup x_n = b \in \mathbb{R} \implies \exists$  subsequence  $x_{n_k} \rightarrow b$ .

**Proof3.** We will construct a subsequence  $n_k$  inductively such that

$$b - \varepsilon_k < x_{n_k} < b + \varepsilon_k, \quad \varepsilon_k = \frac{1}{k}.$$

Given  $\varepsilon = 1$ , by Proposition 1.5.5(1),  $\exists N_1$  s.t.  $n \geq N_1 \implies x_n < b + \varepsilon_1$ . Further, by Proposition 1.5.5(2), for  $M = N_1$ ,  $\exists n_1 > N_1$  s.t.  $x_{n_1} > b - \varepsilon_1$ . Therefore,

$$b - \varepsilon_1 < x_{n_1} < b + \varepsilon_1.$$

**Claim** Given  $k_n$ , we can find  $n_{k+1}$  s.t.  $n_{k+1} > n_k$ , and

$$b - \frac{1}{k+1} < x_{n_{k+1}} < b + \frac{1}{k+1}.$$

After  $\{x_{n_k}\}$  is constructed, use the sandwich lemma to prove  $x_{n_k} \rightarrow b$ .

Q.E.D. ■

**Remark 1.15** Similar arguments hold for  $\liminf x_n = a$ .

**Proposition 1.5.7 Relation Between Cluster Points and Limit:** Let  $x_n \in \mathbb{R}$  be a given sequence.

1. If  $x$  is a cluster point  $\implies \liminf x_n \leq x \leq \limsup x_n$ .
2. If  $a = \liminf x_n$  is finite  $\implies a$  is the smallest cluster point.
3. If  $b = \limsup x_n$  is finite  $\implies b$  is the largest cluster point.
4.  $x_n \rightarrow x \in \mathbb{R} \iff \liminf x_n = \limsup x_n = x$ .

**Proof4.** (of (1)) Suppose  $x$  is a cluster point. Then,  $\exists$  subsequence  $x_{n_k} \rightarrow x$  as  $k \rightarrow \infty$ .

[WTS:  $a_n \leq x \leq b_n \quad \forall n$ ]

For each  $n$ ,  $b_n = \sup \{x_{n+1}, x_{n+2}, \dots\} \geq x_{n_k}$  for large enough  $k$ . Let  $k \rightarrow \infty$ , we have  $b_n \geq x$ . Similarly,  $a_n = \inf \{x_{n+1}, x_{n+2}, \dots\} \leq x_{n_k}$  for large enough  $k$ . As  $k \rightarrow \infty$ ,  $a_n \leq x$ .

So,  $a_n \leq x \leq b_n$ . Take the limit as  $n \rightarrow \infty$ :

$$\lim_{n \rightarrow \infty} a_n \leq x \leq \lim_{n \rightarrow \infty} b_n \implies \liminf x_n \leq x \leq \limsup x_n.$$

Q.E.D. ■

## 1.6 Euclidean Space $\mathbb{R}^n$ and General Metric Space

**Notation 1.1.**  $\mathbb{R}^n = \{(x_1, x_2, \dots, x_n) \mid x_1, \dots, x_n \in \mathbb{R}\}$ .

**Remark 1.16 ( $\mathbb{R}^n$  is a Vector Space)** We can write its standard bases as  $\{e_1, e_2, \dots, e_n\}$ , and the general representation of  $x$  will be

$$x = \sum_{j=1}^n x_j e_j.$$

**Definition 1.6.2 (Norm and Metric).** For  $x, y \in \mathbb{R}^n$ , define *norm* (or length) as

$$\|x\| = \sqrt{\sum_{i=1}^n x_i^2}$$

and the *metric* (distance) as

$$d(x, y) = \|x - y\| = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}.$$

**Definition 1.6.3 (Inner Product).** We define the *inner product* (or dot product) as

$$\langle x, y \rangle = \sum_{i=1}^n x_i y_i.$$

Geometrically, if  $\theta$  is the angle between  $x$  and  $y$ , then

$$\langle x, y \rangle = \|x\| \cdot \|y\| \cdot \cos \theta.$$

So, if  $x \perp y$ ,  $\langle x, y \rangle = 0$ .

**Proposition 1.6.4 Properties of Inner Product:** Suppose  $\langle \cdot, \cdot \rangle$  is an inner product, then

- Positive definite:  $\langle x, x \rangle \geq 0$  and  $\langle x, x \rangle = 0 \iff x = 0$ .
- Linearity:  $\langle x, y + z \rangle = \langle x, y \rangle + \langle x, z \rangle$  and  $\langle \alpha x, y \rangle = \alpha \langle x, y \rangle$ .
- Symmetry:  $\langle x, y \rangle = \langle y, x \rangle$ .

**Proposition 1.6.5 Properties of Norm:** Suppose  $\|\cdot\|$  is a norm, then

- Positive definite:  $\|x\| \geq 0$  and  $\|x\| = 0 \iff x = 0$ .
- Linearity:  $\|\alpha x\| = |\alpha| \cdot \|x\|$ .
- Triangle Inequality:  $\|x + y\| \leq \|x\| + \|y\|$ .

**Proposition 1.6.6 Properties of Metric:** Suppose  $d(\cdot, \cdot)$  is a metric, then

- Positive definite:  $d(x, y) \geq 0$  and  $d(x, y) = 0 \iff x = y$ .
- Symmetry:  $d(x, y) = d(y, x)$ .
- Triangle Inequality:  $d(x, y) \leq d(x, z) + d(z, y)$ .

**Remark 1.17** Inner product always induces a norm. Norm always induced a metric.

**Theorem 1.6.7 Cauchy-Schwarz Inequality**

$$|\langle x, y \rangle| \leq \|x\| \cdot \|y\|.$$

**Example 1.6.8 Use Cauchy-Schwarz Inequality to Prove Triangle Inequality of Norms**

**Proof 1.** Note that

$$\begin{aligned}
 \|x + y\|^2 &= \langle x + y, x + y \rangle && \text{[Definition]} \\
 &= \langle x + y, x \rangle + \langle x + y, y \rangle && \text{[Distribution]} \\
 &= \langle x, x \rangle + \langle y, x \rangle + \langle x, y \rangle + \langle y, y \rangle && \text{[Dsitributionl]} \\
 &= \|x\|^2 + \|y\|^2 + 2\langle x, y \rangle && \text{[Symmetry]} \\
 &\leq \|x\|^2 + \|y\|^2 + 2 \cdot \|x\| \cdot \|y\| && \text{[Cauchy-Schwarz]} \\
 &= (\|x\| + \|y\|)^2.
 \end{aligned}$$

Q.E.D. ■

**Definition 1.6.9 (General Metric Space).** A *metric space*  $(M, d)$  is a set  $M$  and a function  $d : M \times m \rightarrow \mathbb{R}$  s.t.  $\forall x, y, z \in M$ , the following conditions hold:

- Positive definite:  $d(x, y) \geq 0$  and  $d(x, y) = 0 \iff x = y$ .
- Symmetry:  $d(x, y) = d(y, x)$ .
- Triangle Inequality:  $d(x, y) \leq d(x, z) + d(z, y)$ .

**Definition 1.6.10 (General Normed Space).** A *normed space*  $(V, \|\cdot\|)$  is a vector space  $V$  together with a function  $\|\cdot\| : V \rightarrow \mathbb{R}$  s.t.  $\forall x, y \in V$  and  $\forall \alpha \in \mathbb{R}$ ,

- Positive definite:  $\|x\| \geq 0$  and  $\|x\| = 0 \iff x = 0$ .
- Linearity:  $\|\alpha x\| = |\alpha| \cdot \|x\|$
- Triangle Inequality:  $\|x + y\| \leq \|x\| + \|y\|$

**Definition 1.6.11 (General Inner Product Space).** An *inner product space*  $(V, \langle \cdot, \cdot \rangle)$  is a vector space  $V$  and a function  $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}$  s.t.  $\forall x, y, z \in V$  and  $\forall \alpha \in \mathbb{R}$ :

- Positive definite:  $\langle x, x \rangle \geq 0$  and  $\langle x, x \rangle = 0 \iff x = 0$ .
- Symmetry:  $\langle x, y \rangle = \langle y, x \rangle$ .
- Linearity:  $\langle \alpha x, y \rangle = \alpha \langle x, y \rangle$  and  $\langle x + y, z \rangle = \langle x, z \rangle + \langle y, z \rangle$ .

**Example 1.6.12**

- $\mathbb{R}^n$  is a metric space with  $d(x, y) = \|x - y\|$ .
- *Discrete Metric*: Given any set  $M$ , define

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

- *Bounded Metric*: Given metric space  $(M, d)$ , define  $\rho : M \times M \rightarrow \mathbb{R}$ :

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

**Claim 1.6.13**  $(M, \rho)$  is also a metric space.

- $\mathbb{R}^2$  is a metric space under the taxicab metric  $d_1 : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$ :

$$d_1((x_1, y_1), (x_2, y_2)) = |x_2 - x_1| + |y_2 - y_1|.$$

- Let  $\mathcal{C}([0, 1])$  be the collection of all continuous function  $f : [0, 1] \rightarrow \mathbb{R}$ . Define

$$\langle f, g \rangle = \int_0^1 f(x)g(x) \, dx.$$

Then,  $\mathcal{C}$  is an inner product space.

**Remark 1.18 (Relation Among Inner Product, Normed, and Metric Space)**

$$\text{Inner Product} \implies \text{Norm} \implies \text{Metric}$$

- An inner product  $\langle \cdot, \cdot \rangle$  induces a norm:

$$\|x\| = \sqrt{\langle x, x \rangle}.$$

- A norm  $\|\cdot\|$  always induces a metric:

$$d(x, y) = \|x - y\|.$$

**Theorem 1.6.14 General Cauchy-Schwarz Inequality**

In an inner product space  $(V, \langle \cdot, \cdot \rangle)$ , we have  $\forall v, w \in V$ ,

$$|\langle v, w \rangle| \leq \langle v, v \rangle^{\frac{1}{2}} \cdot \langle w, w \rangle^{\frac{1}{2}}.$$

**Proof2.** If  $v = 0$  or  $w = 0$ , it is trivial.

Assume  $v \neq 0$  and  $w \neq 0$ . For any  $t \in \mathbb{R}$ , consider

$$\langle tv + w, tv + w \rangle$$

Then,

$$0 \leq \langle tv + w, tv + w \rangle = t^2 \underbrace{\langle v, v \rangle}_a + 2t \underbrace{\langle v, w \rangle}_b + \underbrace{\langle w, w \rangle}_c$$

Let  $f(t) = at^2 + 2bt + c$  be a 2<sup>nd</sup> order polynomial of  $t$ . Note that  $f(t) \geq 0 \quad \forall t \in \mathbb{R}$ . On the other hand (OTOH), since  $a = \langle v, v \rangle > 0$ ,  $f(t)$  has minimum where  $f'(t) = 0$ .

$$f'(t) = 2at + 2b = 0$$

$$t = -\frac{b}{a}.$$

So,  $f\left(-\frac{b}{a}\right) \geq 0$ , or

$$\left(-\frac{b}{a}\right)^2 a + 2b\left(-\frac{b}{a}\right) + c \geq 0$$

$$\frac{b^2}{a} - 2\frac{b^2}{a} + c \geq 0$$

$$c \geq \frac{b^2}{a}$$

$$b^2 \leq ac$$

$$(\langle v, w \rangle)^2 \leq \langle v, v \rangle \cdot \langle w, w \rangle$$

$$|\langle v, w \rangle| \leq \langle v, v \rangle^{\frac{1}{2}} \cdot \langle w, w \rangle^{\frac{1}{2}}.$$

Q.E.D. ■

## 2 Topology of Euclidean Space

### 2.1 Open Set

**Definition 2.1.1 (Neighborhood & Open Set).** Let  $(M, d)$  be a metric space. Fix  $x \in M$  and  $\varepsilon > 0$ .

- *Neighborhood (nbdd):*

$$D(x, \varepsilon) = \{y \in M \mid d(x, y) < \varepsilon\}.$$

It is also referred as  $\varepsilon$ -nbdd,  $\varepsilon$ -disk, or  $\varepsilon$ -ball.

- *Open Set:* A set  $A \subset M$  is *open* if  $\forall x \in A, \exists \varepsilon > 0$  s.t.  $D(x, \varepsilon) \subset A$ .

#### Example 2.1.2 Open Set

- The unit disk  $D = \{(x, y) \in \mathbb{R}^2 \mid x^2 + y^2 < 1\}$  is open in  $\mathbb{R}^2$ .
- The interval  $(0, 1) \subset \mathbb{R}^1$  is open.
- Given any metric space  $(M, d)$  and  $x_0 \in M$ . The disk

$$D(x_0, r) = \{x \in M \mid d(x, x_0) < r\}$$

is open  $\forall r > 0$ .

**Proof 1.** Fix  $x \in D(x_0, r)$ . [WTS:  $\exists \varepsilon > 0$  s.t.  $D(x, \varepsilon) \subset D(x_0, r)$ .]

Since  $x \in D(x_0, r)$ , by definition,  $d(x, x_0) < r$ . Hence,  $\varepsilon = r - d(x, x_0) > 0$ .

**Claim 2.1.3**  $D(x, \varepsilon) \subset D(x_0, r)$ .

*Proof.* Let  $y \in D(x, \varepsilon)$ . Then,

$$\begin{aligned} d(y, x_0) &\leq d(y, x) + d(x, x_0) \\ &< \varepsilon + d(x, x_0) \\ &= r - \cancel{d(x_0, x)} + \cancel{d(x_0, x)} \\ &= r. \end{aligned}$$

So,  $d(y, x_0) < r$ . By definition,  $y \in D(x_0, r)$ .  $\square$

So,  $D(x, \varepsilon) \subset D(x_0, r)$ . By definition,  $D(x_0, r)$  is open.

Q.E.D.  $\blacksquare$

- The set  $S = \{(x, y) \in \mathbb{R}^2 \mid xy > 1\}$  is open.

**Proof 2.** Given  $(x, y) \in S$ . [WTS:  $\exists \varepsilon > 0$  s.t.  $D((x, y), \varepsilon) \subset S$ .]

Since  $xy > 1$ ,  $\lambda = \frac{1}{2} \left(1 - \frac{1}{xy}\right) > 0$ .

WLOG, assume  $x > 0$  and  $y > 0$ .

Let  $\varepsilon = \min \{\lambda x, \lambda y\}$ . Then, for  $(u, v) \in D((x, y), \varepsilon)$ , we have

$$\begin{aligned} d((u, v), (x, y)) &< \varepsilon \\ \sqrt{(x - u)^2 + (y - v)^2} &< \varepsilon. \end{aligned}$$

So,  $|x - u| < \varepsilon$  and  $|y - v| < \varepsilon$ . Then,

$$\begin{aligned} x \left| 1 - \frac{u}{x} \right| &< \varepsilon \\ \frac{u}{x} &> 1 - \frac{\varepsilon}{x} \geq 1 - \frac{\lambda x}{x} = 1 - \lambda. \end{aligned}$$

Similarly,

$$\frac{v}{y} > 1 - \lambda.$$

Then,

$$\begin{aligned} u \cdot v &= \frac{u}{x} \cdot \frac{v}{y} \cdot (xy) > (1 - \lambda)^2 (xy) \\ &> (1 - 2\lambda)(xy) = 1. \end{aligned}$$

So, as  $uv > 1$ ,  $(u, v) \in S$ . Hence,  $S$  is open.

**Sketch.** Given  $xy > 0$ ; Want  $uv > 1$ . Note that

$$\begin{aligned} uv &= \underbrace{\frac{u}{x}}_{(1-\lambda)} \cdot \underbrace{\frac{v}{y}}_{(1-\lambda)} \cdot xy \\ &= (1 - \lambda)^2 (xy) \\ &> (1 - 2\lambda + \lambda^2)(xy) \\ &> (1 - 2\lambda)(xy) \\ &\geq 1 \\ \implies 1 - 2\lambda &\geq \frac{1}{xy}. \end{aligned}$$

Q.E.D. ■

### Remark 2.1

- In the above definition,  $\varepsilon$  depends on the point  $x$ .
- The open set is defined w.r.t. the underline metric space.



**Example 2.1.4**

$A = (0, 1)$ . Then,  $A$  is an open set as a subset in  $\mathbb{R}^1$ . However,  $A$  is not open as a subset in  $\mathbb{R}^2$ .

**Proposition 2.1.5 Properties of Open Set:** Let  $(M, d)$  be a metric space. Then,

- The intersection of a finite number of open sets is open.
- The union of any number of open sets is open.
- $\emptyset$  and  $M$  are open.

**Proof 3.** (of ①) Suppose  $A = \bigcap_{j=1}^n A_j$ . Fix  $x \in A$ . By definition,  $x \in A_j \quad \forall j = 1, \dots, n$ . Then, we can find  $\varepsilon_j > 0$  s.t.  $D(x, \varepsilon_j) \in A_j$ . As  $A_j$  is open. Take  $\varepsilon = \min \{\varepsilon_1, \varepsilon_2, \dots, \varepsilon_n\}$ . We know

$$D(x, \varepsilon) \in A_j \quad \forall j = 1, \dots, n.$$

Hence,  $D(x, \varepsilon) \in \bigcap_{j=1}^n A_j$ . So,  $A$  is open.

Q.E.D. ■

**Remark 2.2** The intersection of infinitely many number of open sets may not be open.

**Definition 2.1.6 (Interior Point).** Let  $A \subset M$ . A point  $x \in A$  is called an *interior point* of  $A$  if  $\exists \varepsilon > 0$  s.t.  $D(x, \varepsilon) \subset A$ . The *interior* of  $A$  is the collection of all interior points, denoted by  $\text{int}(A)$ .

**Example 2.1.7**

- $A = \{x_0\} \subset \mathbb{R}^n$ ,  $\text{int}(A) = \emptyset$  as there is no nbdd around the point  $x_0$ .
- $A = (0, 1) \subset \mathbb{R}^1$ ,  $\text{int}(A) = A$ .

**Remark 2.3** A set is open if every point in  $A$  is an interior point of  $A$ .

- $B = [0, 1] \subset \mathbb{R}^1$ ,  $\text{int}(B) = (0, 1)$ .

**Proposition 2.1.8 Properties of  $\text{int}(A)$ :**

- $\text{int}(A)$  is open.
- $\text{int}(A)$  is the union of all open subsets of  $A$ .

**Remark 2.4** Or,  $\text{int}(A)$  is the largest open subset of  $A$ .

- $A$  is open  $\iff A = \text{int}(A)$ .

## 2.2 Closed Sets

**Definition 2.2.1 (Closed Set).** A set  $A \subset M$  is *closed* if its complement,  $A^C = M \setminus A$ , is open.

### Example 2.2.2

- $A = [0, 1] \subset \mathbb{R}^1$ .  
 $A^C = (-\infty, 0) \cup (1, +\infty)$ .  
 $A^C$  is open  $\implies A$  is closed.
- $B = \{(x, y) \in \mathbb{R}^2 \mid 1 < x^2 + y^2 \leq 4\}$ .  
 $B^C = \{(x, y) \in \mathbb{R}^2 \mid x^2 + y^2 \leq 1 \text{ or } x^2 + y^2 > 4\}$ .  
 $B$  is not open and not closed.
- A single point set is closed.
- $B(x, \varepsilon) = \{y \in M \mid d(y, x) \leq \varepsilon\}$  is closed.

**Proposition 2.2.3 Basic Properties of Closed Sets:** Given  $(M, d)$ , then

- Union of finite number of closed set is closed.
- Intersection of any number of closed set is closed.
- $\emptyset$  and  $M$  are always closed.

**Remark 2.5** In property ①, one cannot replace “finite number” by “countably many.”

**Definition 2.2.4 (Accumulation Point).** A point  $x \in M$  is an *accumulation point* of the set  $A$  if  $\forall \varepsilon > 0, \exists y \in A$  s.t.  $y \neq x$  and  $y \in D(x, \varepsilon)$ . The collection of accumulation points of  $A$  is denoted as  $\text{ac}(A)$ .

**Remark 2.6**  $x$  does not need to be in  $A$ .

**Definition 2.2.5 (Closure/ $\text{cl}(A)$ ).**

$$\begin{aligned} \text{cl}(A) &= \text{intersection of all closed sets containing } A \\ &= A \cup \text{ac}(A). \end{aligned}$$

**Definition 2.2.6 (Boundary of  $A/\partial A/\text{bd}(A)$ ).**

$$\begin{aligned} \text{bd}(A) &= \partial A = \text{cl}(A) \cap \text{cl}(M \setminus A) \\ &= \text{cl}(A) \setminus \text{int}(A). \end{aligned}$$

**Theorem 2.2.7 Equivalent Conditions of Closed Sets**

Let  $A \subset M$ , the following are equivalent (TFAE):

- $A$  is closed.
- $\text{ac}(A) \subset A$ .
- $A = \text{cl}(A)$ .
- $\text{bd}(A) \subset A$ .

**Proof 1.**

(①  $\implies$  ②): Let  $A \subset M$  be closed and  $x \in \text{ac}(A)$ . [WTS:  $x \in A$ .] Assume  $x \notin A$ . Then,  $x \in M \setminus A$ .

[Proof by contradiction.]

Since  $A$  is closed,  $M \setminus A$  is open, which means  $x \in \text{int}(M \setminus A)$ . That is,  $\exists \varepsilon > 0$  s.t.  $D(x, \varepsilon) \subset M \setminus A$ . Hence,  $D(x, \varepsilon) \cap A = \emptyset$ . \* This contradicts with the assumption that  $x \in \text{ac}(A)$ . As  $D(x, \varepsilon) \cap A = \emptyset$ ,  $\nexists y \in A$  s.t.  $y \in D(x, \varepsilon)$ . Hence,  $x \in A$ .  $\square$

(②  $\iff$  ③): We have  $\text{cl}(A) = A \cup \text{ac}(A)$ .

( $\Rightarrow$ ): If ② is true,  $\text{ac}(A) \subset A$ . Then,  $\text{cl}(A) = A$ .

( $\Leftarrow$ ): If ③ is true  $\text{cl}(A) = A$ . Then,  $A \cup \text{ac}(A) = A$ , so  $\text{ac}(A) \subset A$ .  $\square$

(③  $\implies$  ④): Note that  $\text{bd}(A) = \text{cl}(A) \cap \text{cl}(M \setminus A)$ . Then,  $\text{bd}(A) \subset \text{cl}(A)$ . If  $A = \text{cl}(A)$ , then  $\text{bd}(A) \subset \text{cl}(A) = A$ .  $\square$

(④  $\implies$  ①): Suppose  $\text{bd}(A) \subset A$ . Assume  $A$  is not closed, then  $M \setminus A$  is not open. [Proof by contradiction.] So,  $\exists x_0 \in M \setminus A$  that is not an interior point. Hence,  $\forall \varepsilon > 0$ ,  $D(x_0, \varepsilon) \not\subset M \setminus A$ . So,  $D(x_0, \varepsilon) \cap A \neq \emptyset$ . Hence,  $\exists y \in D(x_0, \varepsilon) \cap A$ . Note that  $x_0 \in M \setminus A$  but  $y \in D(x_0, \varepsilon) \cap A$ . So,  $x_0 \neq y$ . By definition,  $x_0 \in \text{ac}(A)$ . \* As  $x_0 \in \text{ac}(A) \subset \text{bd}(A)$ , but  $x_0 \notin A$ , this contradicts with the assumption that  $\text{bd}(A) \subset A$ . Hence,  $A$  must be closed.

Q.E.D.  $\blacksquare$

**Proposition 2.2.8 :**

- $\text{cl}(A) \cap A = A$ .
- If  $A$  is open, then  $\text{bd}(A) \subset M \setminus A$ .

**Definition 2.2.9 (Limit Point of a Set).** A point  $x \in M$  is called a limit point of  $A$  if  $U \cap A \neq \emptyset$  for every open set  $U$  containing  $x$ .

**Proposition 2.2.10 :**

- If  $x \in \text{ac}(A)$ , then  $x$  is a limit point.
- If  $x$  is a limit point of  $A$  and  $x \notin A$ , then  $x \in \text{ac}(A)$ .
- If  $x$  is a limit point of  $A$ ,  $\exists$  a sequence  $x_n \in A$  with  $x_n \rightarrow x$ .
- $A$  is closed  $\iff A$  contains all of its limit points.

### Summary II: Definitions on Point Set Topology

Let  $M$  be a metric space and  $A \subset M$ .

- $x \in A$  is an *interior point* of  $A$  if  $\exists \varepsilon > 0$  with  $D(x, \varepsilon) \subset A$ .
- $A$  is said to be *open* if every point of  $A$  is an interior point, or equivalently,  $\text{int}(A) = A$ .
- A *neighborhood* of a point  $x$  is any open set  $U$  containing  $x$ .
- $A$  is *closed* if its complement  $M \setminus A$  is open.
- A point  $x \in M$  is an *accumulation point* of  $A$  if  $\forall \varepsilon > 0, \exists y \in A$  with  $y \neq x$  and  $y \in D(x, \varepsilon)$ .
- *Closure* of  $A$ :  $\text{cl}(A) = A \cup \text{ac}(A)$ .
- *Boundary* of  $A$ :  $\partial A = \text{bd}(A) = \text{cl}(A) \cap \text{cl}(A \setminus M) = \text{cl}(A) \setminus \text{int}(A)$ .

## 2.3 Convergence

**Definition 2.3.1 (Convergence of a Sequence).** Let  $(M, d)$  be a metric space. Let  $x_k \in M$  be a sequence and  $x \in M$ . We say that  $x_k$  *converges* to  $x$  (write  $x_k \rightarrow x$ ) if  $\forall \varepsilon > 0, \exists N$  s.t.  $d(x_k, x) < \varepsilon \quad \forall k \geq N$ .

### Theorem 2.3.2 Equivalent Definitions of Convergence

- $x_k \rightarrow x \iff \forall$  open set  $U$  containing  $x, \exists N$  s.t.  $x_k \in U \quad \forall k \geq N$ .

**Remark 2.7** This definition replaces  $\varepsilon$ -neighborhood by an arbitrary neighborhood.

- $x_k \rightarrow x \iff d(x_k, x) \rightarrow 0$ .

### Theorem 2.3.3 Equivalent Definition of Convergence in $\mathbb{R}^n$

In  $\mathbb{R}^n$ , write

$$v_k = (v_k^{(1)}, v_k^{(2)}, \dots, v_k^{(n)}) \quad \text{and} \quad v = (v^{(1)}, v^{(2)}, \dots, v^{(n)}).$$

Then,

$$d(v_k, v)^2 = \|v_k - v\|^2 = \sum_{i=1}^n |v_k^{(i)} - v^{(i)}|^2.$$

Thus,  $v_k \rightarrow v \iff v_k^{(i)} \rightarrow v^{(i)} \quad \forall i = 1, \dots, n$

**Proposition 2.3.4 :** Let  $v_k, w_k \in \mathbb{R}^n$  and  $\lambda_k, \lambda \in \mathbb{R}$  with  $v_k \rightarrow v, w_k \rightarrow w, \lambda_k \rightarrow \lambda$ . Then,

- $v_k + w_k \rightarrow v + w$
- $\lambda v_k \rightarrow \lambda v$
- $\lambda_k v_k \rightarrow \lambda v$

### Theorem 2.3.5 Convergence and Closedness

Let  $(M, d)$  be a metric space and  $A \subset M$ .

- $A$  is closed  $\iff$  for every sequence  $x_k \in A$  that converges in  $M$ , the limit lies in  $A$ .
- $x \in \text{cl}(A) \iff \exists x_k \in A$  s.t.  $x_k \rightarrow x$ .

**Proof 1.** (of ①, sketch):

( $\Rightarrow$ ) Assume  $A \subset M$  is closed. Let  $x_k \in A$  be a sequence with  $x_k \rightarrow x \in M$ . [WTS:  $x \in A$ .] Suppose  $x \notin A$ . Then,  $x \in M \setminus A$ .  $A$  is closed  $\implies M \setminus A$  is open  $\implies \exists \varepsilon > 0$  with  $D(x, \varepsilon) \subset M \setminus A$ . As  $x_k \rightarrow x$ , some  $x_k \in D(x, \varepsilon) \subset M \setminus A$ . \* This contradicts with our assumption that  $x_k \in A$ . So,  $x \in A$ .  $\square$

( $\Leftarrow$ ): Suppose  $x_k \in A$  with  $x_k \rightarrow x \in A$ . Assume  $A \subset M$  is not closed. Then,  $M \setminus A$  is not open  $\implies \exists x \in M \setminus A$  s.t.  $\forall \varepsilon > 0, D(x, \varepsilon) \not\subset M \setminus A$ . For  $\varepsilon = \frac{1}{k}$ ,  $\exists x_k \in D(x, \frac{1}{k}) \cap A$ . Then, \*  $x_k \rightarrow x \notin A$ , contradicting with the assumption  $x_k \rightarrow x \in A$ . Hence,  $A$  must be closed.

Q.E.D. ■

## 2.4 Completeness

**Definition 2.4.1 (Cauchy Sequence).**  $\{x_k\} \in M$  is a *Cauchy sequence* if  $\forall \varepsilon > 0, \exists N$  s.t.  $\forall m, n \geq N, d(x_n, x_m) < \varepsilon$ .

**Definition 2.4.2 (Bounded Sequence).** A sequence  $\{x_k\} \in M$  is *bounded* if  $\exists x_0 \in M$  and  $\exists R > 0$  s.t.

$$d(x_0, x_k) \leq R \quad \forall k.$$

Or,  $x_k \in B(x_0, R) \quad \forall k$ , where  $B(x_0, R)$  denotes a closed ball centered at  $x_0$  with radius  $R$ .

**Definition 2.4.3 (Completeness).**  $(M, d)$  is *complete* if every Cauchy sequence in  $M$  converges.

### Example 2.4.4

- $\mathbb{R}^1$  and  $\mathbb{R}^n$  are complete
- $M = \mathbb{R}^1 \setminus \{0\}$  is not complete. For example,  $x_k = \frac{1}{k}$  does not converge in  $\mathbb{R}^1 \setminus \{0\}$ .
- $\mathbb{Q}$  is not complete.

### Proposition 2.4.5 Basic Properties of Cauchy Sequence:

- Cauchy sequence is always bounded.
- Any converging sequence is always Cauchy.
- If a subsequence of a Cauchy sequence converges, then the original sequence converges.

**Proof 1.** (of ①): Suppose  $\{x_k\}$  is Cauchy sequence. [WTS:  $\exists x_0$  and  $\exists R$  s.t.  $x_k \in B(x_0, R) \quad \forall k$ .]

Then, fix  $\varepsilon = 1$ . By Cauchy sequence,  $\exists N$  s.t.  $m, n \geq N \implies d(x_m, x_n) < \varepsilon = 1$ . Define

$$\begin{aligned} R &= \max \{ \varepsilon, d(x_N, x_1), d(x_N, x_2), \dots, d(x_N, x_{N-1}) \} \\ &= \max \{ 1, d(x_N, x_k) : k = 1, \dots, N-1 \} \end{aligned}$$

Then, we have  $d(x_k, x_N) \leq R \quad \forall k$ , which implies that Cauchy sequence is bounded.

Q.E.D. ■

### Theorem 2.4.6 Closedness and Completeness

Let  $(M, d)$  be a metric space.

- $N \subset M$  is complete  $\implies N$  is closed. [Completeness is stronger than closedness]
- $N \subset M$  is closed and  $M$  is complete  $\implies N$  is complete.

**Remark 2.8** If  $(M, d)$  is a metric space and  $N \subset M$ , then  $(N, d)$  is also a metric space.

**Proof 2.**

- (of ①): Suppose  $N \subset M$  is complete. [WTS: every sequence  $x_k \in N$  that converges, the limit is in  $N$ .]

Given  $\{x_k\} \in N$  with  $x_k \rightarrow x \in M$ . [WTS:  $x \in N$ .]

Since  $\{x_k\} \in M$  converges, it is Cauchy. Further, as  $N \subset M$  is complete, by definition,  $x_k \rightarrow x \in N$  as desired.  $\square$

- (of ②): Suppose  $N \subset M$  is closed and  $M$  is complete. [WTS: Cauchy sequence  $x_k \rightarrow x \in N$ .]

Given  $x_k \in N$  is a Cauchy sequence. Then,  $x_k \in M$  as  $N \subset M$ . Since  $M$  is complete, we know  $x_k \rightarrow x \in M$ . Further, as  $N$  is closed, we know  $x_k \rightarrow x \in N$ . Hence, every Cauchy sequence converges in  $N$ . By definition,  $N$  is complete.

Q.E.D.  $\blacksquare$

**Definition 2.4.7 (Cluster Point).**  $x$  is a *cluster point* of  $\{x_k\}$  if  $\forall \varepsilon > 0$ ,  $\exists$  infinitely many indices  $k$  s.t.  $d(x_k, x) < \varepsilon$ .

**Proposition 2.4.8 Properties of Cluster Points:**

- $x$  is a cluster point  $\iff \forall \varepsilon > 0, \forall N, \exists k > N$  s.t.  $d(x_k, x) < \varepsilon$ .
- $x$  is a cluster point  $\iff \exists$  subsequence  $x_{n_k} \rightarrow x$ .
- $x_k \rightarrow x \iff$  each subsequence  $x_{n_k} \rightarrow x$ .
- $x_k \rightarrow x \iff$  each subsequence has a further subsequence that converges to  $x$ .



### 3 Compactness and Connectedness

#### 3.1 Compactness

**Definition 3.1.1 (Cover and Subcover).** Let  $A \subset M$ .

- A *cover* of a set  $A \subset M$  is a collection  $\{U_i\}$  of sets  $U_i \subset M$  such that

$$\bigcup_i U_i \supset A.$$

- We say  $\{U_i\}$  of  $A$  is an *open cover* if each  $U_i$  is open.
- A *subcover* of a given cover is a subcollection of  $\{U_i\}$  whose union contains  $A$ .
- We say a cover is a *finite cover* if the subcollection contains finite number of sets.

**Example 3.1.2**

Suppose  $A = [0, 1] \subset \mathbb{R}^1$ . Consider

$$U_1 = (-1, 0.1), \quad U_2 = (0, 0.5), \quad U_3 = (0.5, 1).$$

$$U_4 = (0.2, 0.6), \quad U_5 = (0.8, 2), \quad U_6 = (0, 1).$$

Then,

- $\{U_1, \dots, U_6\}$  is a finite cover of  $A$ .
- It is also an open cover.
- $\{U_1, U_5, U_6\}$  is a subcover.

**Definition 3.1.3 (Compactness).** A set  $A \subset M$  is called *compact* if every open cover of  $A$  has a finite subcover.

**Definition 3.1.4 (Sequentially Compact).** A set  $A \subset M$  is *sequentially compact* if every sequence in  $A$  has a subsequence that converges to a point in  $A$ .

**Definition 3.1.5 (Totally Bounded).** A set  $A \subset M$  is *totally bounded* if  $\forall \varepsilon > 0, \exists$  finite set  $\{x_1, x_2, \dots, x_N\} \subset M$  s.t.

$$A \subset \bigcup_{i=1}^N D(x_i, \varepsilon).$$

**Remark 3.1**

- $A$  is sequentially compact  $\implies A$  is closed and bounded.

**Proof 1.** Suppose  $A$  is unbounded. Fix  $x_0 \in M$ . For any  $n \geq 1, \exists x_n \in A$  s.t.

$$d(x_n, x_0) \geq n.$$

By sequential compactness,  $\exists$  subsequence  $x_{n_k} \rightarrow x \in A$  such that

$$\begin{aligned} d(x_{n_k}, x_0) &\leq d(x_{n_k}, x) + d(x, x_0) \\ &< \varepsilon + d(x, x_0). \end{aligned}$$

Take  $\varepsilon = 1$ ,  $d(x_{n_k}, x_0) < 1 + d(x, x_0)$  is a finite number. However,  $d(x_{n_k}, x_0) \geq n_k$ .  $\times$  As  $n_k \rightarrow \infty$ ,  $1 + d(x, x_0)$  is a finite number, we reach a contradiction. Hence,  $A$  must be bounded.

Q.E.D. ■

- $A$  is totally bounded  $\implies A$  is bounded.

**Theorem 3.1.6 Bolzano-Weirstrass Theorem (B-W Thm.)**

$A \subset M$  is compact  $\iff A$  is sequentially compact.

**Proof2.**

**Lemma 3.1.7:**  $A \subset M$  is compact  $\implies A$  is closed.

*Proof.* [WTS:  $M \setminus A$  is open.]

Fix  $x \in M \setminus A$ . For  $n = 1, 2, \dots$ , let  $U_n = \left\{ y \mid d(x, y) > \frac{1}{n} \right\}$ .

**Claim**  $\{U_n \mid n = 1, 2, \dots\}$  is an open cover of  $A$ .

*Proof.* In fact, let  $a \in A$ . Then,  $d(a, x) > 0$ . By Archimedean,  $\exists n$  s.t.

$$\frac{1}{n} < d(a, x).$$

This implies that  $a \in U_n$ . So,  $a \in \bigcup_{i=1}^{\infty} U_i$ . That is,  $A \subset \bigcup_{i=1}^{\infty} U_i$ .  $\square$

By the compactness,  $\exists$  finite subcover, say  $\{U_1, \dots, U_N\}$ . Thus,

$$A \subset \bigcup_{i=1}^N U_i = U_N = \left\{ y \mid d(y, x) > \frac{1}{N} \right\}.$$

Therefore,

$$D\left(x, \frac{1}{N}\right) = \left\{ y \mid d(y, x) < \frac{1}{N} \right\} \subset M \setminus A.$$

Hence, by definition,  $M \setminus A$  is open, and so  $A$  must be closed.  $\square$

**Lemma 3.1.8 (When is the converse of Lemma 3.1.7 true?):**

$B \subset M$  is closed and  $M$  is compact  $\implies B$  is compact.

*Proof.* Given an open cover  $\{V_i \mid i \in I\}$  of  $B$ . [WTS:  $\exists$  a finite subcover of  $B$ .]

Since  $B$  is closed,  $M \setminus B$  is open. Then,

$$\{V_i \mid i \in I\} \cup \{M \setminus B\} \text{ is an open cover of } M.$$

Since  $M$  is compact,  $\exists$  a finite subcover of  $M$ :

$$\{V_1, V_2, \dots, V_N\} \cup \{M \setminus B\}.$$

Note that

$$\bigcup_{i=1}^N V_i \supset B,$$

we know

$$\{V_1, V_2, \dots, V_N\} \text{ is a finite subcover of } B.$$

Hence, by definition,  $B$  is compact.  $\square$

( $\Rightarrow$ ): Now, we prove the forward direction of the B-W Theorem. Let  $A \subset M$  be compact. [WTS:  $A$  is sequentially compact]

- Set Up: Given a sequence  $\{x_k\} \in A$ . [WTS:  $\exists x_{n_k} \rightarrow x \in A$ ]

By Lemma 3.1.7, compactness  $\implies$  closedness. Since  $A$  is closed, all converging sequence converges to some point in  $A$ . Hence, we only need to show  $\exists$  converging subsequence.

- Reduction: To this end, we may assume that  $\{x_k\}$  contains a subsequence of distinct terms. Denote this subsequence by  $\{y_k\}$ . [WTS:  $\{y_k\}$  has a convergent subsequence]

If  $\{x_k\}$  does not contain subsequence of distinct terms, then  $\{x_k\}$  is a constant sequence after sufficient terms. Therefore, it must converge and is trivial in this discussion.

- Suppose, for the sake of contradiction,  $\{y_k\}$  does not have a convergent subsequence.
- **Claim**  $y_k$ 's are "isolated:" For each  $k = 1, 2, \dots, \exists$  neighborhood  $U_k$  of  $y_k$  s.t.  $y_j \notin U_k$  for any  $j \neq k$ .

*Proof.* Suppose, for the sake of contradiction, that the claim does not hold. Then,  $\exists k$  with the property  $\forall \varepsilon > 0, \exists j \neq k$  s.t.  $y_j \in U_k = D(y_k, \varepsilon)$ . Take  $\varepsilon = \frac{1}{m}$ . We obtain subsequence  $y_{j_m} \in D\left(y_k, \frac{1}{m}\right), m = 1, 2, \dots$ . Hence, when  $m \rightarrow \infty, y_{j_m} \rightarrow y_k$ .

This implies  $\{y_k\}$  has a convergent subsequence.  $\times$  This contradicts with our assumption that  $\{y_k\}$  does not have a convergent subsequence. Hence, the claim must be true.  $\square$

- Now, proceed with the assumption that this claim is true.

Consider the set formed by elements in  $\{y_k\}$ :

$$B = \{y_1, y_2, \dots\}$$

Since  $\{y_n\}$  has no convergent subsequence,  $B$  has no accumulation point, and so  $\text{cl}(B) = B$ , which implies  $B$  is closed.

By Lemma 3.1.8,  $B$  is compact.

On the other hand,  $\{U_k\}$  is an open cover of  $B$ . But by claim,  $\exists$  no finite subcover. \* This contradicts with the fact that  $B$  is compact. Thus,  $\{y_k\}$  has a convergent subsequence, which converges to a point because  $A$  is closed.

□

( $\Leftarrow$ ): Now, let's consider the backward direction. Suppose  $A \subset M$  is sequentially compact. [WTS:  $A$  is compact]

Let  $\{u_i\}$  be an open cover of  $A$ . [WTS:  $\exists$  a finite subcover]

**Claim (1)**  $\exists r > 0$  s.t. for each  $y \in A$ ,  $D(y, r) \subset U_i$  for some  $i$ .  $\implies$  Each point has a neighborhood of fixed size that is contained in some  $U_i$ .

*Proof.* Suppose otherwise. Then,

$$\forall r = \frac{1}{n} > 0, \exists y_n \in A \text{ s.t. } D\left(y_n, \frac{1}{n}\right) \text{ is not contained in any } U_i.$$

By assumption,  $A$  is sequentially compact. Then,  $\{y_n\}$  has a convergent subsequence  $z_n \rightarrow z \in A$ .

On the other hand,  $U_i$  is an open cover of  $A$ , then  $z_n \in U_{i_0}$  for some  $i_0$ . Further, since  $U_{i_0}$  is open,  $\exists \varepsilon > 0$  s.t.  $D(z, \varepsilon) \subset U_{i_0}$ .

Fix large  $N$  s.t.

$$d(z_N, z) < \frac{\varepsilon}{2}.$$

So,

$$D\left(z, \frac{\varepsilon}{2}\right) \subset D(z, \varepsilon) \subset U_{i_0}.$$

\* This is a contradiction with our assumption that  $D\left(y_n, \frac{1}{n}\right)$  is not contained in any  $U_i$ . Hence, the original claim is true. □

**Claim (2)**  $A$  is totally bounded.

*Proof.* Suppose otherwise. Then,  $\exists \varepsilon > 0$  s.t.  $A$  cannot be covered by finite number of balls of radius  $\varepsilon$ . Choose  $y_1 \in A$  and  $y_2 \in A \setminus D(y_1, \varepsilon)$ . Then, choose  $y_3 \in A \setminus (D(y_1, \varepsilon) \cup D(y_2, \varepsilon))$ . This process can go forever as  $A$  cannot be covered by finite number of balls of radius  $\varepsilon$ . So, we get sequence

$$y_n \in A \setminus (D(y_1, \varepsilon) \cup \dots \cup D(y_{n-1}, \varepsilon)).$$

We have a sequence  $\{y_n\}$  with the property that

$$d(x_n, x_m) > \varepsilon \quad \forall n \neq m.$$

So,  $\{y_n\}$  does not have a convergent subsequence.

Everything convergent must be Cauchy.  $d(x_n, x_m) > \varepsilon$  implies not Cauchy, so it must be non-convergent. \* This contradicts with the assumption that  $A$  is sequentially compact (has a subsequence converges to some point in  $A$ ). Hence, this claim must be true. □

Now, let  $r > 0$  be as in Claim (1). By Claim (2),  $\exists y_1, y_2, \dots, y_N \in A$  s.t.

$$A \subset \bigcup_{j=1}^N D(y_j, r).$$

Then, further by Claim (1), we get  $D(y_j, r) \subset U_{i_j}$ . So,

$$A \subset \bigcup_{j=1}^N D(y_j, r) \subset \bigcup_{j=1}^N U_{i_j}.$$

Therefore,  $A$  can be covered by a finite subcover. Hence,  $A$  is compact.

Q.E.D. ■

### Theorem 3.1.9

$A \subset M$  is compact  $\iff A$  is complete and totally bounded.

**Remark 3.2** So, if a set is not bounded/totally bounded, it cannot be compact.

**Proof 3.** ( $\Rightarrow$ ): Done when proving B-W Thm. □

( $\Leftarrow$ ): Assume  $A$  is complete and totally bounded. [WTS:  $A$  is compact/sequentially compact]

Let  $\{y_n\}$  be a sequence in  $A$ . [WTS:  $\exists$  subsequence  $y_{n_k}$  converges in  $A$ ]

WLOG, we may assume  $\{y_n\}$  is formed by distinct terms. If we don't get distinct terms, we will have a constant sequence when  $n$  gets sufficiently large. Hence, it converges in  $A$  and is trivial to discuss.

Since  $A$  is totally bounded, for  $\varepsilon_1 = 1$ ,  $A$  is covered by finite number of balls:

$$D(x_1^{(1)}, \varepsilon_1), \dots, D(x_{L_1}^{(1)}, \varepsilon_1).$$

We can choose a subsequence  $\{y_{1n}\}_{n=1}^\infty$  of  $\{y_n\}$  that is contained one of the balls.

Repeat that for  $\varepsilon_2 = \frac{1}{2}$ , we have

$$A \subset D(x_1^{(2)}, \varepsilon_2) \cup \dots \cup D(x_{L_2}^{(2)}, \varepsilon_2).$$

We can choose a subsequence  $\{y_{2n}\}_{n=1}^\infty$  of  $\{y_n\}$  that is contained in one of the balls.

Continuing this process with  $\varepsilon_m = \frac{1}{m}$ ,  $m = 1, 2, \dots$ . We obtain a subsequence  $\{y_{mn}\}_{n=1}^\infty$  that is contained in a ball of radius  $\varepsilon_m = \frac{1}{m}$ . Then, we have the following subsequence:

$$\begin{array}{ccccccc} y_{11}, & y_{12}, & y_{13}, & \cdots, & y_{1n}, & \cdots \\ y_{21}, & y_{22}, & y_{23}, & \cdots, & y_{2n}, & \cdots \\ \vdots & & & & & \\ y_{m1}, & y_{m2}, & y_{m3}, & \cdots, & y_{mn}, & \cdots \\ \vdots & & & & & \end{array}$$

Each subsequence is a subsubsequence of the proceeding subsequence.

Select  $y_{11}, y_{22}, y_{33}, \dots, y_{nn}, \dots$  to form a subsequence of  $\{y_n\}$ .

Denote this subsequence as  $\{z_n\} = \{y_{nn}\}$ .

$A$  is complete. To show  $z_n$  converge in  $A$ , we only need to show  $z_n$  is Cauchy.

**Claim**  $\{z_n\}$  is Cauchy.

*Proof.* Assume  $n > m$ :

$$d(z_n, z_m) < \frac{2}{m}.$$

When  $m \rightarrow \infty$ ,  $d(z_n, z_m) \rightarrow 0$ . So,  $\{z_n\}$  is Cauchy.  $\square$

Since  $A$  is complete,  $\{z_n\}$  is Cauchy, we have  $z_n \rightarrow z \in A$ . Hence,  $A$  is sequentially compact. By B-W Theorem,  $A$  is compact.

Q.E.D.  $\blacksquare$

### 3.2 Compactness in $\mathbb{R}^n$

#### Theorem 3.2.1 Heine-Borel Theorem

A set  $A \subset \mathbb{R}^n$  is compact  $\iff A$  is bounded and closed.

**Proof 1.**  $(\implies)$ : True in general metric space.  $\square$

$(\impliedby)$ : Assume  $A \subset \mathbb{R}^n$  is closed and bounded. [WTS:  $A$  is sequentially compact]

Given sequence  $\{x_k\}$  in  $A$ , write

$$x_k = (x_k^{(1)}, x_k^{(2)}, \dots, x_k^{(n)}) \in A \subset \mathbb{R}^n.$$

$A$  is bounded  $\implies \{x_k\}$  is bounded  $\implies \{x_k^{(1)}\}$  is bounded in  $\mathbb{R}$ .

$\implies \exists$  converging subsequence  $\{x_{f_1(k)}^{(1)}\}_{k=1}^\infty$ .

Similarly,  $\{x_{f_1(k)}^{(2)}\}_{k=1}^\infty$  is bounded in  $\mathbb{R}$ .  $\implies \exists$  converging subsequence  $\{x_{f_2(k)}^{(2)}\}_{k=1}^\infty$ .

In this way, we obtain subsequence

$$x_{f_n(k)} = (x_{f_n(k)}^{(1)}, x_{f_n(k)}^{(2)}, \dots, x_{f_n(k)}^{(n)})$$

with  $x_{f_n(k)}^{(i)} \xrightarrow{k \rightarrow \infty} x^{(i)}$  for  $i = 1, 2, \dots, n$ . Hence,

$$x_{f_n(k)} \rightarrow (x^{(1)}, x^{(2)}, \dots, x^{(n)}) \in A.$$

Therefore,  $A$  is sequentially compact.

Q.E.D.  $\blacksquare$

**Remark 3.3** In Heine-Borel Theorem,  $(\impliedby)$  does not hold in general metric space. That is,  $A$  metric space that is closed and bounded does not imply compactness. For example, let  $M =$  infinite set with discrete

*metric*

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

*M is closed and bounded, but M is not compact.*

### Example 3.2.2

- $A \subset \mathbb{R}^n$  is bounded  $\implies \text{cl}(A)$  is compact.
- $A = [0, 1] \subset \mathbb{R}^1$  is compact.
- $A = (0, 1] \subset \mathbb{R}$  is not compact.
- $\mathbb{R}$  is not compact because it is not totally bounded.

## 3.3 Nested Set Property

### Theorem 3.3.1 Nested Set Property

Let  $F_k$  be a set of non-empty compact sets in  $M$  s.t.

$$F_{k+1} \subset F_k \quad \forall k = 1, 2, \dots$$

Then,

$$\bigcap_{k=1}^{\infty} F_k \neq \emptyset.$$

**Proof 1.** For each  $k = 1, 2, \dots$ , choose  $x_k \in F_k$ . Then,  $\{x_k\} \subset F_1$ . Since  $F_1$  is compact,  $\exists$  subsequence

$$x_{f(k)} \xrightarrow{k \rightarrow \infty} x \in F_1.$$

**Claim**  $x \in F_n \quad \forall n$ .

**Proof.** Fix  $n > 1$ . Then, for large  $k$  ( $\exists N$  s.t.  $k \geq N$ ), we have  $f(k) \geq n$ . Then,  $F_{f(k)} \subset F_n$ . Recall that  $x_{f(k)} \in F_{f(k)}$  and  $x_{f(k)} \xrightarrow{k \rightarrow \infty} x$ , then

$$x \in F_n$$

as  $F_n$  is closed.  $\square$

Hence,  $x \in \bigcap_{k=1}^{\infty} F_k \neq \emptyset$ .

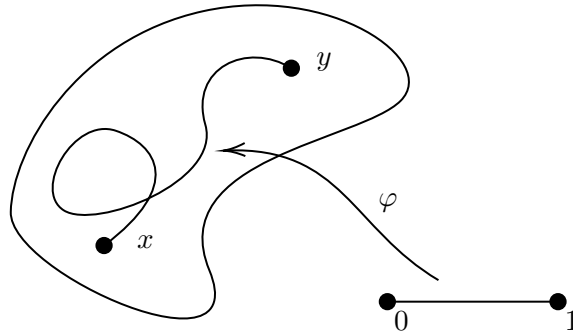
Q.E.D.  $\blacksquare$

**Remark 3.4** “Compact” cannot be replaced by “open,” “closed,” or “bounded open.”

### 3.4 Connectedness

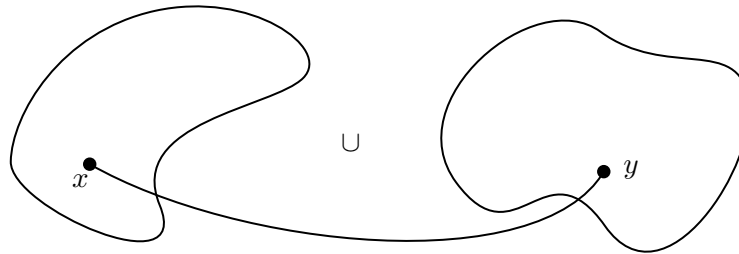
**Definition 3.4.1 (Path-Connected, Geometric Point of View).** A set  $A \subset M$  is *path-connected* if each pair of points  $x, y \in A$  can be joined by a continuous path given by a continuous map

$$\varphi : [0, 1] \rightarrow A \quad \text{s.t.} \quad \varphi(0) = x \quad \text{and} \quad \varphi(1) = y$$



#### Example 3.4.2

- This is not path-connected:

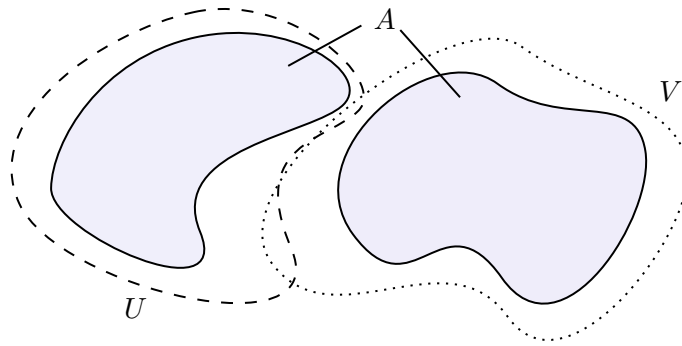


- Let  $\varphi : [0, 1] \rightarrow M$  be a continuous map. Then,  $C = \varphi([0, 1]) \subset M$  (the image of this mapping) is path-connected.

**Definition 3.4.3 (Disconnected Set, Topological Point of View).** A set  $A \subset M$  is said to be *disconnected* if  $\exists$  open sets  $U, V \subset M$  that separate A:

- $U \cap V \cap A = \emptyset$
- $U \cap A \neq \emptyset$  and  $V \cap A \neq \emptyset$
- $A \subset U \cup V$





**Definition 3.4.4 (Connected Set).** If a set is not disconnected, then it is *connected*.

**Remark 3.5** It is easy to prove disconnectedness since we only need to find one pair of open sets satisfying the 4 conditions. To prove connectedness, we need to show  $\forall$  open sets  $U, V \subset M$ , they cannot satisfy the 4 conditions at the same time.

**Theorem 3.4.5**

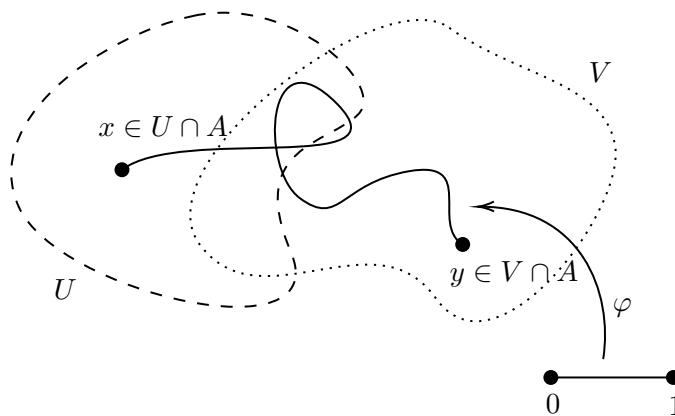
Path-connectedness  $\implies$  connectedness

**Proof 1.** We will start the proof with the following claim (The proof is trivial, and so we omit the proof):

**Claim 3.4.6** The interval  $[a, b] \subset \mathbb{R}^1$  is connected.

Suppose, for the sake of contradiction, that  $A \subset M$  is path-connected but not connected. Then,  $\exists$  open sets  $U, V$  that separates  $A$  as defined in Definition 3.4.3.

Fix  $x \in U \cap A$  and  $y \in V \cap A$ .



Since  $A$  is path-connected,  $\exists$  a continuous map  $\varphi : [0, 1] \rightarrow A$  with  $\varphi(0) = x$  and  $\varphi(1) = y$ . Let

$$\begin{aligned} C &= \varphi^{-1}(A \cap U) \subset [0, 1] \\ &:= \{t \in [0, 1] \mid \varphi(t) \in A \cap U\}. \end{aligned}$$

Similarly, we can define  $D = \varphi^{-1}(A \cap V)$ . Then,  $0 \in C$  and  $1 \in D$ .

**Claim 3.4.7**  $C$  is closed.

*Proof.* Let  $t_k \in C$  s.t.  $t_k \rightarrow t$ . Then, by continuity of  $\varphi$ ,  $\varphi(t_k) \rightarrow \varphi(t) \in A$ . Suppose, for the sake of contradiction,  $\varphi(t) \notin U$ . Then,  $\varphi(t) \in V$ . Since  $V$  is open,  $\varphi(t_k) \in V$  for large  $k$ . Hence,

$$\varphi(t_k) \in A \cap U \cap V = \emptyset.$$

✱ We reach a contradiction. So,  $\varphi(t) \in U$ , which implies  $t \in C$ . As  $t_k \rightarrow t \in C$ , we have shown that  $C$  is closed.  $\square$

**Corollary 3.4.8 :** By symmetry of  $C$  and  $D$ ,  $D$  is also closed.

To derive a contradiction with Claim 3.4.6, note that

$$A \cap U \cap V = \emptyset,$$

which implies  $C \cap D = \emptyset$ . Therefore, the two open sets  $(\mathbb{R} \setminus C)$  and  $(\mathbb{R} \setminus D)$  separates  $[0, 1]$ . ✱ This contradicts with Claim 3.4.6 that  $[0, 1]$  is connected. Hence, our assumption was wrong, and  $A$  must be path-connected and connected. In other words, path-connectedness  $\implies$  connectedness.

Q.E.D.  $\blacksquare$

**Remark 3.6** *The converse is not true.*

**Example 3.4.9**

$$\text{Suppose } A = \underbrace{\left\{ \left( x, \sin \frac{1}{x} \right) \mid x > 0 \right\}}_{\text{graph of } f(x) = \sin\left(\frac{1}{x}\right)} \cup \underbrace{\{(0, y) \mid -1 \leq y \leq 1\}}_{\text{segment of } y\text{-axis}} \subset \mathbb{R}^2.$$

Then,  $A$  is connected but not path-connected.

**Proposition 3.4.10 :**  $A \subset \mathbb{R}^n$  open and connected  $\implies$  path-connected.

**Proof2.** (Sketch) Fix a point  $x_0 \in A$  s.t.

$$B = \{y \in A \mid x_0 \text{ and } y \text{ can be joined by a continuous path } \in A\}.$$

Show:

- $B \neq \emptyset$  [ $x_0 \in B$ ]
- $B$  is open.
- $B$  is closed in  $A$ .

Then,  $B = A$ . [If  $B \neq A$ , then  $U = B$  and  $V = A \setminus B$  separates  $A \implies A$  is disconnected  $\implies$  contradiction, it must be  $A = B$ .]

Q.E.D.  $\blacksquare$

**Theorem 3.4.11 Equivalent Ways to Describe Connectedness**

- In Definition 3.4.3, one can replace “open sets” by “closed sets.”

$A \subset M$  is disconnected  $\iff \exists$  closed sets  $E, F$  s.t.

- $E \cap F \cap A = \emptyset$
- $E \cap A \neq \emptyset$  and  $F \cap A \neq \emptyset$
- $A \subset E \cup F$

[Take complement of open sets, we get closed sets]

- In Definition 3.4.3, one can replace “ $U, V$ ” by “disjoint open sets.”

$A \subset M$  is disconnected  $\iff \exists$  disjoint open sets  $U_1$  and  $V_1$  s.t.

- $U_1 \cap V_1 \cap A = \emptyset$
- $U_1 \cap A \neq \emptyset$  and  $V_1 \cap A \neq \emptyset$
- $A \subset U_1 \cup V_1$

**Proof3.** (Hint of ②): Consider the distance function  $d(x, A \cap V)$  given fixed  $x \in U \cap A$ .

**Claim**  $\forall x \in A \cap U$ , *define*  $d(x) = d(x, A \cap V) = \inf \{d(x, a) \mid a \in A \cap V\}$ . *Then*,  $d(x) > 0$ . *Similarly*,  $\forall y \in A \cap V$ , *define*  $d(y) = d(y, A \cap U) = \inf \{d(y, a) \mid a \in A \cap U\}$ . *Then*,  $d(y) > 0$ .

Define open sets  $U_1, V_1$  as follows:

$$U_1 = \left\{ D\left(x, \frac{1}{2}d(x)\right) \mid x \in A \cap U \right\} \quad \text{and} \quad V_1 = \left\{ D\left(y, \frac{1}{2}d(y)\right) \mid y \in A \cap V \right\}$$

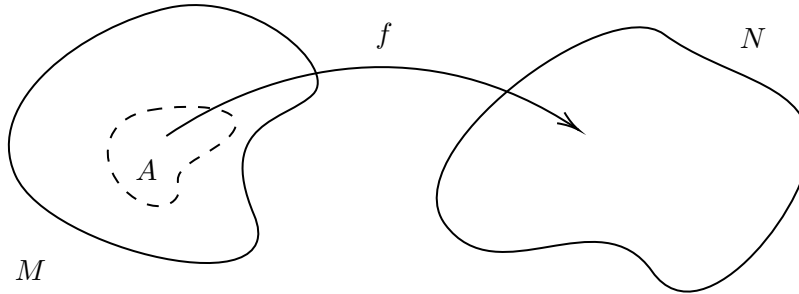
We have the desired disjoint  $U_1$  and  $V_1$ .

Q.E.D. ■

## 4 Continuous Mappings

### 4.1 Continuity

**Definition 4.1.1 (Maps).** Suppose  $(M, d)$  and  $(N, \rho)$  are metric spaces. Let  $A \subset M$ . Then,  $f : A \rightarrow N$  is a *map* (or a function)



**Definition 4.1.2 (Continuous Maps).**  $f$  is *continuous at a point*  $x_0 \in A$  if

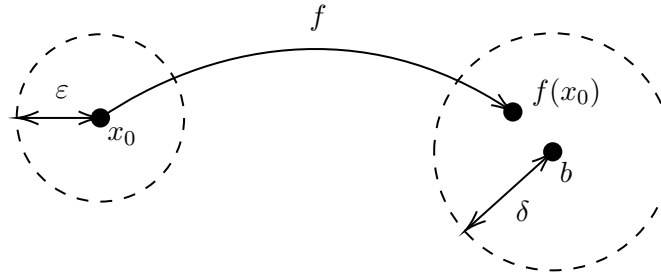
$$\lim_{\substack{x \rightarrow x_0 \\ x \in A}} f(x) = f(x_0).$$

$f$  is *continuous in A* if it is continuous at each point in  $A$ .

**Definition 4.1.3 (Limit of a Function).**  $b \in N$  is the limit of  $f(x)$  at  $x_0$ , written as

$$\lim_{x \rightarrow x_0} f(x) = b,$$

if  $\forall \varepsilon > 0, \exists \delta > 0$  s.t.  $x \in A$  and  $d(x, x_0) < \delta \implies \rho(f(x), b) < \varepsilon$ .



**Definition 4.1.4 (Isolated Points).**  $x_0 \in A$  is an *isolated point* in  $A$  if  $\exists \delta > 0$  s.t.  $D(x_0, \delta) \cap A = \{x_0\}$ .

**Remark 4.1**

- The continuous definition implies three things: the function is defined, the limit exists, and the limit value equals the function value.
- A point is either an isolated point or an accumulation point.
- For the limit definition,  $x_0$  is not required to be in  $A$ . For example,

$$f(x) = \frac{\sin(x)}{x}, \quad x \in (0, 1) \quad \lim_{x \rightarrow 0} f(x) = 0 \notin (0, 1).$$

- If  $x_0$  is an isolated point in  $A$ , then  $\lim_{x \rightarrow x_0} f(x) = f(x_0)$  is always true. Therefore, any function  $f(x)$  is continuous at isolated points.

**Example 4.1.5**

- $f(x) = x : \mathbb{R}^n \rightarrow \mathbb{R}^n$  (identity function) is continuous
- $g(x) = \begin{cases} x, & 0 \leq x \leq 1 \\ 2x, & 1 < x \leq 3 \end{cases} : [0, 3] \rightarrow \mathbb{R}^1$  is continuous at every point except for  $x = 1$ .
- $h(x) = \begin{cases} x, & x \neq 1 \\ 3, & x = 1 \end{cases} : \mathbb{R} \rightarrow \mathbb{R}$  is continuous at every point except  $x = 1$ .

**Theorem 4.1.6 Equivalent Conditions for Continuity**

Let  $f : A \subset M \rightarrow N$ . The following are equivalent:

- $f$  is continuous on  $A$ .
- For each converging sequence  $x_k \rightarrow x \in A$ ,  $f(x_k) \rightarrow f(x)$ .

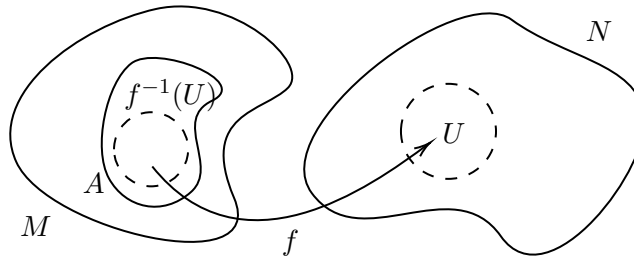
**Remark 4.2** Continuous map preserves the convergence of sequences

- For each open set  $U \subset N$ , the pre-image  $f^{-1}(U) \subset A$  is open relative to  $A$ . That is

$$f^{-1}(U) = \{x \in A \mid f(x) \in U\} = A \cap V, \quad \text{where } V \subset M \text{ is open.}$$

- For each close set  $F \subset N$ , the pre-image  $f^{-1}(F) \subset A$  is closed relative to  $A$ . That is,

$$f^{-1}(F) = A \cap E, \quad \text{where } E \subset M \text{ is closed.}$$



**Proof 1.** We will prove equivalence by the following cycle: ①  $\implies$  ②  $\implies$  ④  $\implies$  ③  $\implies$  ①.

(①  $\implies$  ②): Given sequence  $x_k \in A$  with  $x_k \rightarrow x \in A$ . [WTS:  $\lim_{k \rightarrow \infty} f(x_k) = f(x)$ ]

(②  $\implies$  ④): Fix closed set  $F \in N$ . [WTS:  $f^{-1}(F) = A \cap \text{cl}(f^{-1}(F))$ ] It is trivial that  $f^{-1}(F) \subset A \cap \text{cl}(f^{-1}(F))$ . So, we only need to prove the “ $\supset$ ” direction. Given  $x \in A \cap \text{cl}(f^{-1}(F))$ ,  $\exists$  sequence  $x_n \in f^{-1}(F) \subset A$  s.t.  $x_n \rightarrow x$ . Then,  $y_n = f(x_n) \rightarrow f(x) \in F$  by ② and closedness. So,  $x \in f^{-1}(F)$ . That is,  $A \cap \text{cl}(f^{-1}(F)) \supset f^{-1}(F)$ . Hence,  $f^{-1}(F) = A \cap \text{cl}(f^{-1}(F))$ , implying  $f^{-1}(F)$  is closed in  $A$ .

(④  $\implies$  ③): [Use complement:  $U \subset N$  is open  $\iff F = N \setminus U$  is closed]

(③  $\implies$  ①): Given  $x_0 \in A$ . [WTS:  $\lim_{x \rightarrow x_0} f(x) = f(x_0)$ ] Fix any  $\varepsilon > 0$ . [WTS:  $\exists \delta > 0$  s.t.  $d(x, x_0) < \delta \implies \rho(f(x), f(x_0)) < \varepsilon$ ] Let  $U = D(f(x_0), \varepsilon) \subset M$  is open. By ③,  $f^{-1}(U)$  is open in  $A$ . i.e.,

$$f^{-1}(U) = A \cap V, \quad V \subset M \text{ is open.}$$

Note that  $x_0 \in f^{-1}(U) \implies x_0 \in V$ . Since  $V$  is open,  $\exists \delta > 0$  s.t.  $D(x, \delta) \subset V$ . [WTS:  $x \in A$ ,  $d(x, x_0) < \delta \implies \rho(f(x), f(x_0)) < \varepsilon$ ] Suppose  $x \in A$  with  $d(x, x_0) < \delta$ . Then,  $x \in A$  and  $x \in V$ . That is,  $x \in A \cap V = f^{-1}(U)$ . Hence,  $f(x) \in U$ . By definition of  $U$ , we get  $\rho(f(x), f(x_0)) < \varepsilon$  as desired.

Q.E.D. ■

## 4.2 Properties of Continuous Mappings

### Theorem 4.2.1 Images of Compact and Connected Sets

Suppose  $f : M \rightarrow N$  is continuous. Then,

- If  $K \subset M$  is compact, then  $f(K)$  is also compact.
- If  $B \subset M$  is connected, then  $f(B)$  is also connected.

**Proof 1.**

- Let  $x_k$  be a sequence in  $K$ . Then,  $y_k = f(x_k)$  is a sequence in  $f(K)$ . [WTS:  $f(K)$  is sequentially compact.] Suppose  $K$  is compact,  $\exists x_{k_j} \rightarrow x_0 \in K$  when  $j \rightarrow \infty$ . By continuity of  $f$ ,  $f(x_{k_j}) \rightarrow f(x_0) \in f(K)$  when  $k \rightarrow \infty$ . So, for sequence  $y_k = f(x_k)$ , we find a subsequence  $f(x_{k_j}) \rightarrow f(x_0) \in f(K)$ . So,  $f(K)$  is sequentially compact. □
- Given connected set  $B \subset M$ . Assume, for the sake of contradiction, that  $f(B)$  is disconnected. Then,  $\exists$  open sets  $U, V$  s.t.  $f(B) \cap U \cap V = \emptyset$  and  $f(B) \cap U \neq \emptyset$ ,  $f(B) \cap V \neq \emptyset$ ,  $f(B) \subset U \cup V$ . [We can derive that  $B$  is also disconnected, which is a contradiction.] So, it must be that  $f(B)$  is also connected.

Q.E.D. ■

### Theorem 4.2.2 Operations on Continuous Mapping

Addition, multiplication, divisions, and compositions of continuous functions (if they are well-defined) are also continuous.

**Example 4.2.3**

If  $f(x) = \mathbb{R} \rightarrow \mathbb{R}$ ,  $g : \mathbb{R} \rightarrow \mathbb{R}$  are continuous, then,  $f(x)g(x)$  is also continuous.

**Proof 2.** Denote  $F(x) = f(x)g(x)$ . Then,

$$\begin{aligned}
 |F(x) - F(x_0)| &= |f(x)g(x) - f(x_0)g(x_0)| \\
 &\leq |f(x)g(x) - f(x)g(x_0)| + |f(x)g(x_0) - f(x_0)g(x_0)| \\
 &= |f(x)||g(x) - g(x_0)| + |g(x_0)||f(x) - f(x_0)| \\
 &\quad \vdots \\
 &< \varepsilon
 \end{aligned}$$

Q.E.D. ■

**Theorem 4.2.4 Maximum/Minimum Property**

Let  $K \subset M$  be compact and  $f : K \rightarrow \mathbb{R}$  be continuous. Then,

- $f$  is bounded on  $K$  (i.e.,  $f(K)$  is a bounded set)
- $\exists x_0, x_1 \in K$  s.t.

$$f(x_1) = \max_{x \in K} f(x) \quad \text{and} \quad f(x_0) = \min_{x \in K} f(x).$$

That is,  $f(x_0) \leq f(x) \leq f(x_1) \quad \forall x \in K$ .

**Proof 3.**

- Since  $K$  is compact and  $f$  is continuous,  $f(K)$  is compact. Since  $f(K) \subset \mathbb{R}$  is compact,  $f(K)$  is closed and bounded.
- Since  $f(K)$  is bounded, we know  $\inf(f(K))$  and  $\sup(f(K))$  exist and are finite. Further since  $f(K)$  is closed,  $\inf(f(K), \sup(f(K)) \in f(K)$ . Hence,  $\exists x_0 = \inf(f(K))$  and  $x_1 = \sup(f(K))$  s.t.

$$f(x_0) \leq f(x) \leq f(x_1) \quad \forall x \in K.$$

Q.E.D. ■

**Remark 4.3**

- The condition “compact” cannot be removed.

**Example 4.2.5**

$$f(x) = \frac{1}{x} : (0, 1) \rightarrow \mathbb{R} \text{ is continuous but not bounded}$$

$f(x) = x : (0, 1) \rightarrow \mathbb{R}$  is bounded, but does not have max/min values

- The condition “continuity” cannot be removed.

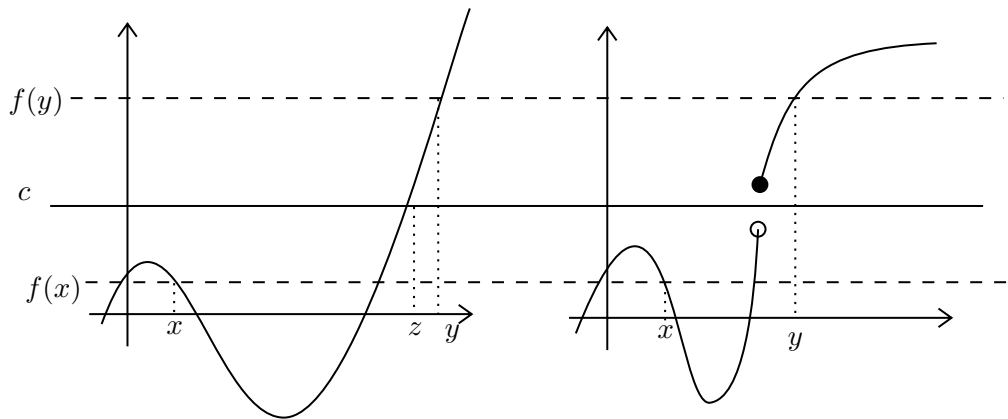
**Example 4.2.6**

Consider function  $f : [0, 1] \rightarrow \mathbb{R}$  by  $f(x) = \begin{cases} \frac{1}{x}, & x > 0 \\ 2, & x = 0. \end{cases}$  Although  $[0, 1]$  is compact,  $f(x)$  is not continuous, and  $f$  is not bounded and does not have max/min values on  $[0, 1]$ .

- We don't need differentiability here.

**Theorem 4.2.7 Intermediate Value Theorem (IVT)**

Let  $K \subset M$  be connected and  $f : K \rightarrow \mathbb{R}$  be continuous. Suppose  $x, y \in K$  with  $f(x) < f(y)$ . Then, for any intermediate value  $c$  s.t.  $f(x) < c < f(y)$ ,  $\exists z \in K$  with  $x < z < y$  s.t.  $f(z) = c$ .



**Proof 4.** Let  $K \subset M$  be connected and  $f : K \rightarrow \mathbb{R}$  be continuous. Suppose  $x, y \in K$  with  $f(x) < f(y)$ . Assume, for the sake of contradiction,  $\exists c$  with  $f(x) < c < f(y)$  s.t.  $c \notin f(K)$ .

Since  $K$  is connected and  $f$  is continuous,  $f(K)$  is also connected. However,  $U = (-\infty, c)$  and  $V = (c, +\infty)$  separate  $f(K)$ , implying  $f(K)$  is not connected. ✱ We reach a contradiction. So, such a  $c$  does not exist.

Q.E.D. ■

**Example 4.2.8 Application of IVT I**

Let  $f(x)$  be a polynomial of odd degree. Then,  $f$  has at least one real root.

**Proof 5.** Suppose  $f(x) : \mathbb{R} \rightarrow \mathbb{R}$  is continuous. Write  $f(x)$  as

$$f(x) = a_n x^n + a_{n-1} x^{n-1} + \cdots + a_1 x + a_0,$$



where  $a_n \neq 0$  and  $n = 2k + 1$  is odd.

WLOG, suppose  $a_n > 0$ . Then,

$$\lim_{x \rightarrow -\infty} f(x) = -\infty \quad \text{and} \quad \lim_{x \rightarrow \infty} f(x) = \infty.$$

So,  $\exists x, y \in \mathbb{R}$  s.t.  $f(x) < 0$  and  $f(y) > 0$ . Therefore, by IVT,  $\exists x_0 \in \mathbb{R}$  s.t.  $f(x_0) = c = 0$ .

Q.E.D. ■

**Definition 4.2.9 (Fixed Point).**  $x$  is a *fixed point* of  $f$  if  $f(x) = x$ .

**Example 4.2.10 Application of IVT II**

Let  $f : [1, 2] \rightarrow [0, 3]$  be continuous with  $f(1) = 0$ ,  $f(2) = 3$ . Then,  $f$  has a fixed point.

**Proof 6.** Apply IVT to a new function:  $F(x) = f(x) - x$ . Take  $c = 0$  as the intermediate value.

Q.E.D. ■

### 4.3 Uniform Continuity (UC)

**Definition 4.3.1 (Uniform Continuity (UC)).** A function  $f : A \subset M \rightarrow N$  is *uniformly continuous* on  $A$  if  $\forall \varepsilon > 0, \exists \delta > 0$  s.t.  $x, y \in A$  and  $d(x, y) < \delta \implies \rho(f(x), f(y)) < \varepsilon$ .

**Remark 4.4**

- For uniform continuity, the  $\delta$  depends only on  $\varepsilon$  not on points.
- For continuity (at  $x_0$ ), the  $\delta$  may depend on  $\varepsilon$  and the point  $x_0$ .

**Example 4.3.2**

Consider  $f(x) = \frac{1}{x} : (0, 1) \rightarrow \mathbb{R}$ .  $f$  is continuous at any point  $x_0 \in (0, 1)$ . But to satisfy

$$|x - x_0| < \delta \implies |f(x) - f(x_0)| = \left| \frac{1}{x} - \frac{1}{x_0} \right| = \frac{|x - x_0|}{|x \cdot x_0|} < \varepsilon$$

we need to pick

$$|x - x_0| = \delta = \min \left\{ \frac{1}{2} x_0^2 \varepsilon, \frac{1}{2} x_0 \right\}.$$

**Theorem 4.3.3 Uniform Continuity on Compact Set**

Let  $f : K \subset M \rightarrow N$  be continuous and  $K$  be compact. Then,  $f$  is uniformly continuous on  $K$ .

**Proof 1.** Fix  $\varepsilon > 0$ . For each  $x \in K$ , since  $f$  is continuous at  $x$ ,  $\exists \delta_x$  s.t. for  $y \in K$  with  $d(x, y) < \delta_x$ , we have  $\rho(f(x), f(y)) < \frac{\varepsilon}{2}$ .

Consider the open cover of  $K$ :  $\left\{ D\left(x, \frac{\delta_x}{2}\right) \mid x \in K \right\}$ . Since  $K$  is compact,  $\exists$  subcover:

$$D\left(x_i, \frac{\delta_{x_i}}{2}\right), \quad i = 1, 2, \dots, L$$

Finally, let

$$\delta = \min_{1 \leq i \leq L} \left\{ \frac{\delta_{x_i}}{2} \right\}.$$

**Claim**  $x, y \in K$  with  $d(x, y) < \delta \implies \rho(f(x), f(y)) < \varepsilon$ .

*Proof.* Note that

$$\begin{aligned} d(y, x_i) &\leq d(y, x) + d(x, x_i) \\ &< \delta + \frac{\delta_{x_i}}{2} \\ &< \delta_{x_i}. \end{aligned}$$

One can continue to show that  $\rho(f(x), f(y)) < \varepsilon$ .  $\square$

Q.E.D.  $\blacksquare$

**Definition 4.3.4 (Lipschitz Continuity).** A function  $f : A \subset M \rightarrow N$  is called *Lipschitz* if  $\exists$  constant  $L$  s.t.

$$\rho(f(x), f(y)) \leq L \cdot d(x, y) \quad \forall x, y \in A.$$

#### Theorem 4.3.5 Lipschitz and Uniform Continuity

If  $f : A \subset M \rightarrow N$  is Lipschitz, then  $f$  is uniformly continuous in  $A$ .

**Corollary 4.3.6 :** Suppose  $f : (a, b) \rightarrow \mathbb{R}$  is differentiable and  $\exists M > 0$  s.t.  $|f'(x)| \leq M \quad \forall x \in (a, b)$ . Then,  $f$  is Lipschitz.

**Proof2.** Given  $x, y \in (a, b)$ . Then,

$$\begin{aligned} |f(y) - f(x)| &= |f'(z)(y - x)| && \text{[Mean Value Theorem]} \\ &\leq M|y - x|. \end{aligned}$$

Q.E.D.  $\blacksquare$

#### Example 4.3.7 Lipschitz Functions

$f(x) = x$  and  $f(x) = \sin x$  are Lipschitz functions.

#### Remark 4.5

- If  $f$  has bounded derivative (or slope), then  $f$  is uniformly continuous.

- But if  $f$  is differentiable and uniformly continuous,  $f$  may not have bounded derivative.
- Open End-ed Questions:
  - $f : \mathbb{R} \rightarrow \mathbb{R}$  is bounded and continuous,  $f$  may not be uniformly continuous.
  - $f, g : \mathbb{R} \rightarrow \mathbb{R}$  are uniformly continuous,  $f \cdot g$  is not uniformly continuous in general.
  - But if  $f$ , or  $g$ , or both are bounded and uniformly continuous, is  $f \cdot g$  uniformly continuous?

## 4.4 Differentiability

**Remark 4.6** Starting from this section, we will only consider functions  $f : \text{an interval} \rightarrow \mathbb{R}$ .

**Definition 4.4.1 (Differentiability).** A function  $f$  is *differentiable* at a point  $x_0$  if it is defined in an open interval that contains  $x_0$  and its derivative exists:

$$f'(x_0) = \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0}, \quad (\text{D})$$

or equivalently, set  $h = x - x_0$ ,

$$f'(x_0) = \lim_{h \rightarrow 0} \frac{f(x_0 + h) - f(x_0)}{h}.$$

**Remark 4.7 (Interpretation)**

- Rewrite (D) as

$$\lim_{x \rightarrow x_0} \left[ \frac{f(x) - f(x_0) - f'(x_0)(x - x_0)}{x - x_0} \right] = 0.$$

This implies the function  $y = f(x)$  can be approximated by the linear function

$$y = f(x_0) + f'(x_0)(x - x_0)$$

in a neighborhood of  $x_0$ .

- Rewrite (D) as

$$\lim_{\Delta x \rightarrow 0} \left[ \frac{f(x_0 + \Delta x) - f(x_0)}{\Delta x} - f'(x_0) \right] = 0.$$

this implies the slope of tangent line is the limit of the slope of secant lines.

### Theorem 4.4.2 Continuity of Differentiable Functions

Suppose  $f : A \subset M \rightarrow N$  is differentiable at  $x_0$ . Then, it is continuous at  $x_0$ .

**Proof 1.** Given  $\varepsilon > 0$ . Since

$$\lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0} = f'(x_0),$$

$$\exists \delta_1 > 0 \text{ s.t. } |x - x_0| < \delta_1 \implies \left| \frac{f(x) - f(x_0)}{x - x_0} \right| < |f'(x)| + 1. \text{ Choose}$$

$$\delta = \min \left\{ \frac{\varepsilon}{|f'(x)| + 1}, \delta_1 \right\}.$$

So, when  $|x - x_0| < \delta$ , we have

$$\begin{aligned} |f(x) - f(x_0)| &= \frac{|f(x) - f(x_0)|}{|x - x_0|} \cdot |x - x_0| \\ &< (|f'(x)| + 1) \cdot \frac{\varepsilon}{|f'(x)| + 1} \\ &= \varepsilon. \end{aligned}$$

Q.E.D. ■

**Remark 4.8** *The converse is not true: continuity  $\not\Rightarrow$  differentiability. Counterexample:  $f(x) = |x|$ .*

**Proof2.** (Another Approach) Note that

$$\begin{aligned} \lim_{x \rightarrow x_0} (f(x) - f(x_0)) &= \lim_{x \rightarrow x_0} \left[ \frac{f(x) - f(x_0)}{x - x_0} \right] (x - x_0) \\ &= \lim_{x \rightarrow x_0} \left[ \frac{f(x) - f(x_0)}{x - x_0} \right] \lim_{x \rightarrow x_0} (x - x_0) \quad [\text{Product Rule of Limit}] \\ &= f'(x) \cdot 0 \\ &= 0. \end{aligned}$$

So, the function is continuous.

Q.E.D. ■

#### Theorem 4.4.3 Rules of Differentiation

- Constant multiple rule:

$$(kf)'(x_0) = k \cdot f'(x_0).$$

- Sum rule:

$$(f + g)'(x_0) = f'(x_0) + g'(x_0)$$

- Product rule:

$$(fg)'(x_0) = f'(x_0)g(x_0) + f(x_0)g'(x_0).$$

- Quotient rule:

$$\left( \frac{f}{g} \right)'(x_0) = \frac{f'(x_0)g(x_0) - f(x_0)g'(x_0)}{(g(x_0))^2}$$

- Chain rule:

$$(g \circ f)'(x_0) = g'(f(x_0)) \cdot f'(x_0).$$

**Lemma 4.4.4:** If  $f : (a, b) \rightarrow \mathbb{R}$  is differentiable and  $f$  has a max (or min) at  $c \in (a, b)$ , then  $f'(c) = 0$ .

**Proof3.** Assume  $f$  has a max at  $c \in (a, b)$ . Then,

$$f'(c) = \lim_{h \rightarrow 0} \frac{f(h+c) - f(c)}{h}.$$

[WTS:  $f'(c) \geq 0$  and  $f'(c) \leq 0$ .]

As  $f$  has a max at  $c$ ,  $f(h+c) \leq f(c)$ , and so

$$f(h+c) - f(c) \leq 0.$$

Case I  $h > 0$ :

$$f'(c) = \lim_{h \rightarrow 0^+} \frac{f(h+c) - f(c)}{h} \leq 0.$$

Case II  $h < 0$ :

$$f'(c) = \lim_{h \rightarrow 0^-} \frac{f(h+c) - f(c)}{h} \geq 0.$$

As  $f'(c) \geq 0$  and  $f'(c) \leq 0$ , it must be that  $f'(c) = 0$ .

Q.E.D. ■

#### Theorem 4.4.5 Rolle's Theorem

Let  $f : [a, b] \rightarrow \mathbb{R}$  be continuous and  $f$  be differentiable on  $(a, b)$ . If  $f(a) = f(b) = 0$ , then  $\exists c \in (a, b)$  s.t.  $f'(c) = 0$ .

**Proof4.**  $f$  has max and min on  $[a, b]$  as  $[a, b]$  is compact. [WTS: This max/min occur in  $(a, b)$ .]

Since  $f(a) = f(b) = 0$ , then max and min cannot both occur at the endpoint (i.e., either max or min occur in  $(a, b)$ ) unless  $f$  is the constant function  $f(x) = 0$ .

Now, by Lemma 4.4.4,  $\exists c \in (a, b)$  s.t.  $f'(c) = 0$ , where  $c$  is either the max or min.

Q.E.D. ■

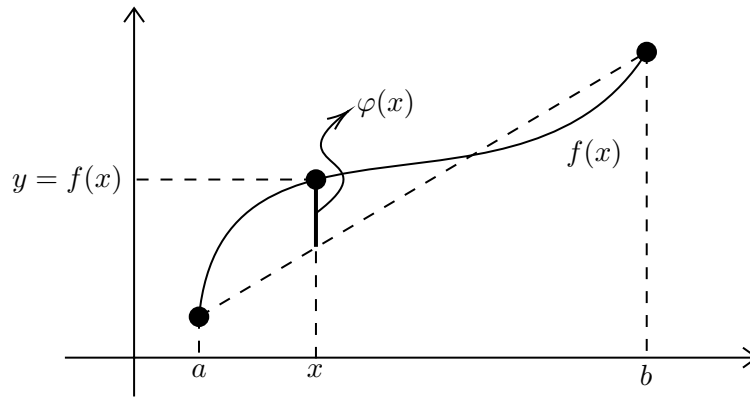
#### Theorem 4.4.6 Mean Value Theorem (MVT)

Suppose  $f$  is continuous on  $[a, b]$  and differentiable on  $(a, b)$ . Then,  $\exists c \in (a, b)$  s.t.

$$f(b) - f(a) = f'(c)(b - a) \quad \text{or} \quad f'(c) = \frac{f(b) - f(a)}{b - a}.$$

**Remark 4.9** Rolle's Theorem is a special case of MVT. We will use the special case to prove the general case.

**Proof5.**



Construct  $\varphi(x)$ :

$$\varphi(x) = f(x) - \left[ f(a) + \frac{f(b) - f(a)}{b - a}(x - a) \right].$$

One can verify the following:

- $\varphi(a) = 0$ ;
- $\varphi(b) = 0$ ; and
- $\varphi$  is continuous and differentiable.

Then, apply Rolle's Theorem to  $\varphi(x)$ :  $\exists c \in (a, b)$  s.t.

$$\varphi'(c) = 0.$$

Note that  $\varphi'(c) = f'(c) - \frac{f(b) - f(a)}{b - a} = 0$ , we have

$$f'(c) = \frac{f(b) - f(a)}{b - a}.$$

Q.E.D. ■

**Remark 4.10 (Geometric Interpretation)** *There is at least one point where the instant change of rate is the same as the average change of rate.*

**Definition 4.4.7 (Monotonicity).**

- We say  $f(x)$  is *increasing* (or *strictly increasing*) at a point  $x_0$  if  $\exists$  open interval  $(a, b)$  containing  $x_0$  with:
  - $a < x < x_0 \implies f(x) \leq f(x_0)$  (or  $f(x) < f(x_0)$ );
  - $x_0 < x < b \implies f(x) \geq f(x_0)$  (or  $f(x) > f(x_0)$ ).
- Similar definition for decreasing (or strictly decreasing) at a point  $x_0$ .
- $f(x)$  is increasing (or strictly increasing) on an interval  $I$  if for  $x_1, x_2 \in I$

$$x_1 < x_2 \implies f(x_1) \leq f(x_2) \quad (\text{or } f(x_1) < f(x_2)).$$

- Similar definition for decreasing (or strictly decreasing) on an interval.

**Theorem 4.4.8 Local Monotonicity and Derivative**

Let  $f$  be differentiable at  $x_0$ . Then,

- $f$  increasing at  $x_0 \implies f'(x_0) \geq 0$ ;  $f$  decreasing at  $x_0 \implies f'(x_0) \leq 0$ .
- $f'(x_0) > 0 \implies f$  strictly increasing at  $x_0$ ;  $f'(x_0) < 0 \implies f$  strictly decreasing at  $x_0$ .

**Proof 6.** (of ①): Suppose  $f$  is increasing at  $x_0$ . Then

$$\begin{aligned} f(x_0 + h) - f(x_0) &\geq 0 \quad \text{when } h > 0 \\ &\leq 0 \quad \text{when } h < 0. \end{aligned}$$

Then,

$$f'(x_0) = \lim_{h \rightarrow 0} \frac{f(x_0 + h) - f(x_0)}{h} \geq 0.$$

(of ②): Suppose  $f'(x_0) > 0$ . Then, for  $\varepsilon = \frac{1}{2}f'(x_0) > 0$ ,  $\exists \delta > 0$  s.t.

$$0 < |h| < \delta \implies \left| \frac{f(x_0 + h) - f(x_0)}{h} - f'(x_0) \right| < \varepsilon = \frac{1}{2}f'(x_0).$$

$$-\frac{1}{2}f'(x_0) < \frac{f(x_0 + h) - f(x_0)}{h} - f'(x_0) < \frac{1}{2}f'(x_0) \implies 0 < \frac{1}{2}f'(x_0) < \frac{f(x_0 + h) - f(x_0)}{h} < \frac{3}{2}f'(x_0).$$

When  $x < x_0$ ,  $h = x - x_0 < 0$ . As  $\frac{f(x_0 + h) - f(x_0)}{h} > 0$ ,

$$f(x_0 + h) - f(x_0) = f(x) - f(x_0) < 0 \implies f(x) < f(x_0)$$

When  $x > x_0$ ,  $h = x - x_0 > 0$ ,

$$f(x_0 + h) - f(x_0) = f(x) - f(x_0) > 0 \implies f(x) > f(x_0).$$

Hence,  $f$  is strictly increasing.

Q.E.D. ■

**Theorem 4.4.9 Global Monotonicity and Derivative**

Let  $f$  be continuous on  $[a, b]$  and differentiable on  $(a, b)$ . Then,

- $f'(x) \geq 0 \quad \forall x \in (a, b) \implies f$  increasing on  $[a, b]$ .
- $f'(x) \leq 0 \quad \forall x \in (a, b) \implies f$  decreasing on  $[a, b]$ .
- $f'(x) > 0 \quad \forall x \in (a, b) \implies f$  strictly increasing on  $[a, b]$ .
- $f'(x) < 0 \quad \forall x \in (a, b) \implies f$  strictly decreasing on  $[a, b]$ .

**Theorem 4.4.10 Local Max/Min and Derivative**

Suppose  $f$  is continuous on  $[a, b]$  and twice differentiable on  $(a, b)$ . Let  $x_0 \in (a, b)$ .

- $f'(x_0) = 0$  and  $f''(x_0) > 0 \implies x_0$  is a strict local min of  $f$ .
- $f'(x_0) = 0$  and  $f''(x_0) < 0 \implies x_0$  is a strict local max of  $f$ .

**Proof 7.** (of ①) By Theorem 3.3.8(2),  $f''(x_0) > 0 \implies f'(x)$  is strictly increasing at  $x_0$ . Then,

- $f'(x) < f'(x_0) = 0 \quad \forall x \in (x_0 - \delta, x_0) \implies f(x)$  strictly decreasing on  $(x_0 - \delta, x_0)$
- $f'(x) > f'(x_0) = 0 \quad \forall x \in (x_0, x_0 + \delta) \implies f(x)$  strictly increasing on  $(x_0, x_0 + \delta)$ .

Q.E.D. ■

**Theorem 4.4.11 Inverse Function Theorem (IFT)**

Suppose  $f'(x) > 0 \quad \forall x \in (a, b)$  (or,  $f'(x) < 0 \quad \forall x \in (a, b)$ ). Then,

- $f : (a, b) \rightarrow \mathbb{R}$  is a bijection onto its image
- Inverse  $f^{-1}$  is differentiable on its domain.
- $(f^{-1})'(y) = \frac{1}{f'(x)}$ , where  $y = f(x)$ .

**Proof 8.** Assume  $f'(x) > 0 \quad \forall x \in (a, b)$ . Then,  $f$  is strictly increasing. Then,  $f$  is 1-to-1 function  $\implies f$  is a bijection  $\implies f^{-1}$  exists. [WTS:  $f^{-1}$  is continuous.]

Let  $U$  be an open set in  $(a, b)$ . [WTS:  $(f^{-1})^{-1}(U) = f(U)$  is open.]



Finally, write  $y = f(x)$ . Then,  $x = f^{-1}(y)$ . Let  $y_0 = f(x_0)$ . Then,

$$\begin{aligned}
 (f^{-1})'(y_0) &= \lim_{y \rightarrow y_0} \frac{f^{-1}(y) - f^{-1}(y_0)}{y - y_0} \\
 &= \lim_{y \rightarrow y_0} \frac{x - x_0}{f(x) - f(x_0)} \\
 &= \lim_{x \rightarrow x_0} \frac{1}{\frac{f(x) - f(x_0)}{x - x_0}} \\
 &= \frac{1}{\lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0}} \\
 &= \frac{1}{f'(x_0)}.
 \end{aligned}$$

Q.E.D. ■

## 4.5 Integration

**Definition 4.5.1 (Riemann Integrable).** Let  $A \subset \mathbb{R}$  be bounded and  $f : A \rightarrow \mathbb{R}$  be a bounded function.

[We want to make sense  $\int_A f(x)dx$ .]

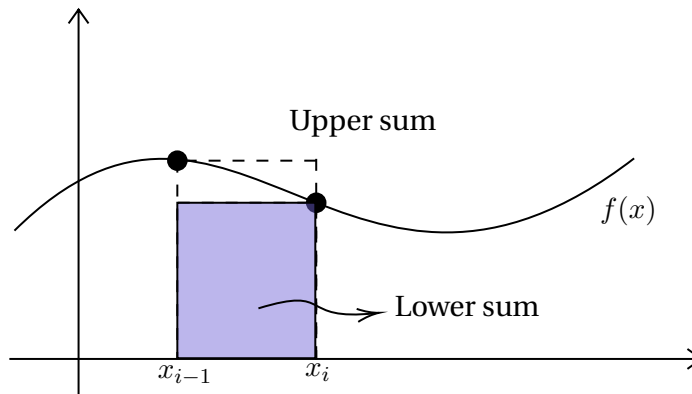
- Partition the interval:

If interval  $[a, b] \supset A$  and extend function  $f(x)$  to  $[a, b]$  by letting  $f(x) = 0 \quad \forall x \notin A$ .

Partition the interval  $[a, b]$  by points:  $a = x_0 < x_1 < \cdots < x_n = b$ . Denote  $P$  by

$$P = \{x_0, x_1, x_2, \dots, x_n\}.$$

- Form Upper and Lower Sum of  $f$ .



For any fixed partition, let

$$U(f, P) = \sum_{i=1}^n \sup \{f(x) \mid x \in [x_{i-1}, x_i]\} (x_i - x_{i-1})$$

is the upper sum, and

$$L(f, P) = \sum_{i=1}^n \inf \{f(x) \mid x \in [x_{i-1}, x_i]\} (x_i - x_{i-1})$$

is the lower sum.

**Claim** Suppose  $m \leq f(x) \leq M$ . Then,

$$m(b-a) \leq L(f, P) \leq U(f, P) \leq M(b-a).$$

- Upper integral and Lower integral are defined as

$$\int_A^{\bar{}} f = \inf \{U(f, P) : P \text{ is a partition}\} \quad (\text{Upper Integral})$$

$$\int_A^{\underline{}} f = \sup \{L(f, P) : P \text{ is a partition}\} \quad (\text{Lower Integral})$$

- We say a function  $f$  is *Riemann integrable* if

$$\int_A^{\bar{}} f = \int_A^{\underline{}} f,$$

and we write

$$\int_A f = \int_A^{\bar{}} f = \int_A^{\underline{}} f.$$

#### Example 4.5.2 Riemann Integrable

- Define  $f : [0, 1] \rightarrow \mathbb{R}$  by

$$f(x) = \begin{cases} 0 & \text{if } x \text{ is rational} \\ 1 & \text{if } x \text{ is irrational.} \end{cases}$$

Then, for any partition  $P$ ,

$$U(f, P) = \sum_{i=1}^n 1 \cdot (x_i - x_{i-1}) = 1$$

and

$$L(f, P) = \sum_{i=1}^n 0 \cdot (x_i - x_{i-1}) = 0.$$

So,

$$\int_A^{\bar{}} f \neq \int_A^{\underline{}} f \implies f \text{ is not integrable}$$

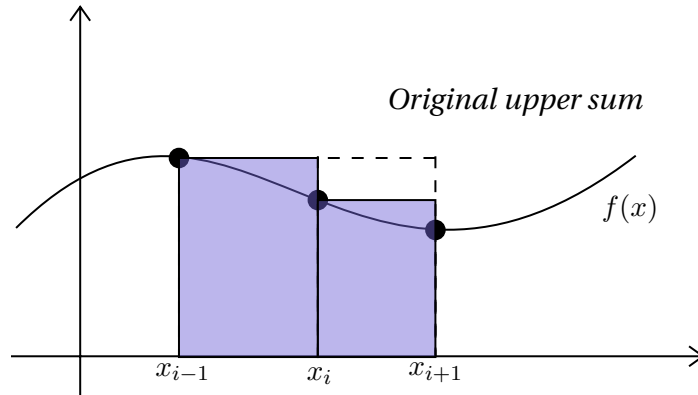
- Compute  $\int_0^1 x \, dx$  and  $\int_0^1 x \, dx$ .

*Hint: Consider partition  $P_n = \left\{ 0 < \frac{1}{n} < \frac{2}{n} < \dots < \frac{n}{n} \right\}$ .*

**Lemma 4.5.3 :** Let  $f : [a, b] \rightarrow \mathbb{R}$  be bounded. If  $P, P'$  are partitions of  $[a, b]$  with  $P \subset P'$  ( $P'$  is a refinement of  $P$ ), then

$$L(f, P) \leq L(f, P') \leq U(f, P') \leq U(f, P).$$

**Remark 4.11** *In words, when the partition gets finer, lower sum increases but upper sum decreases.*



**Proposition 4.5.4 :**

$$\int_a^b f \leq \int_a^b \bar{f}$$

**Proof1.** For any fixed partition  $P$  and  $Q$ . As  $P \subset P \cup Q$  and  $Q \subset P \cup Q$ , by Lemma 4.5.4, we have

$$L(f, P) \leq L(f, P \cup Q) \leq U(f, P \cup Q) \leq U(f, Q).$$

Then,

$$\int_a^b f = \sup_P L(f, P) \leq U(f, Q) \quad \text{for any } Q$$

So,

$$\int_a^b f \leq \inf_Q U(f, Q) = \int_a^b \bar{f}.$$

Q.E.D. ■

#### Theorem 4.5.5

- If  $f : [a, b] \rightarrow \mathbb{R}$  is bounded and is continuous at all but finite many points, then  $f$  is integrable.
- If  $f$  is increasing or decreasing on  $[a, b]$ , then  $f$  is integrable.

**Proof2.**

- (Proof of ①): Observe that  $\forall$  partition  $P$ ,  $L(f, P) \leq \int_a^b f \leq \int_a^b \bar{f} \leq U(f, P)$ . [To prove a function is integrable, it's sufficient to show that  $\forall \varepsilon > 0$ ,  $\exists$  partition  $P$  s.t.  $U(f, P) - L(f, P) < \varepsilon$ .]

- Suppose  $f$  is continuous on  $[a, b]$  except at  $a_1, a_2, \dots, a_k$ . Since  $f$  is bounded,  $\exists m, M$  s.t.  $m \leq f(x) \leq M \quad \forall x \in [a, b]$ . Choose partition  $P_1$  s.t. each subinterval containing some  $a_i$  has length  $\leq \frac{\varepsilon}{2} \cdot \frac{1}{2k(M-m)}$ .

Let  $K$  be the union of the remaining subinterval in  $P_1$ . Then,  $K$  is compact and  $f$  is continuous on  $K$ . So,  $f$  is uniformly continuous on  $K$ . That is,

$$\exists \delta > 0 \text{ s.t. } x_1, x_2 \in K \text{ s.t. } |x_1 - x_2| < \delta \implies |f(x_1) - f(x_2)| < \frac{\varepsilon}{2(b-a)}.$$

- Construct the refinement  $P$  of  $P_1$  so that each subinterval in  $P$  not containing some  $a_i$  has length  $< \delta$ . So,

$$P = \{a = x_0 < x_1 < \dots < x_n = b\} \quad \text{and} \quad I_j = [x_{j-1}, x_j].$$

Denote

$$M_j = \sup_{I_j} f(x) \quad \text{and} \quad m_j = \inf_{I_j} f(x).$$

If  $I_j$  contains some  $a_i$ , then  $m \leq m_j \leq M_j \leq M$ .

If  $I_j$  contains no discontinuous points, then  $I_j \subset K$ , and

$$M_j - m_j = \max - \min < \frac{\varepsilon}{2(b-a)}.$$

- Finally, we have

$$\begin{aligned} U(f, P) - L(f, P) &= \sum_{j=1}^n (M_j - m_j)(x_j - x_{j-1}) \\ &= \sum_{a_i \in I_j} (M_j - m_j)(x_j - x_{j-1}) + \sum_{a_i \notin I_j} (M_j - m_j)(x_j - x_{j-1}) \\ &< \underbrace{2k}_{\text{worse case: } 2k \text{ such intervals}} \underbrace{\frac{(M-m)}{2}}_{\text{estimate of } M_j - m_j} \cdot \underbrace{\frac{\varepsilon}{2} \cdot \frac{1}{2k(M-m)}}_{\text{length of } I_j} + \underbrace{\frac{\varepsilon}{2(b-a)}}_{\text{estimate of } M_j - m_j} \underbrace{(b-a)}_{\text{total length}} \\ &= \frac{\varepsilon}{2} + \frac{\varepsilon}{2} \\ &= \varepsilon. \end{aligned}$$

Therefore,

$$\int_a^b f = \int_a^b \bar{f} \implies f \text{ is integrable.}$$

- (Proof of ②): Assume  $f$  is increasing. Given  $\varepsilon > 0$ . Consider an equal partition

$$P_n = \left\{ a = x_0, x_1 = x_0 + \frac{b-a}{n}, x_2, \dots, x_n = b \right\}.$$

Then, by equal partition and  $f$  is increasing, we have

$$U(f, P_n) = \sum_{j=1}^n f(x_j)(x_j - x_{j-1}) = \frac{b-a}{n} \sum_{j=1}^n f(x_j)$$

and

$$L(f, P_n) = \sum_{j=1}^n f(x_{j-1})(x_j - x_{j-1}) = \frac{b-a}{n} \sum_{j=1}^n f(x_{j-1}).$$

So,

$$\begin{aligned} U(f, P_n) - L(f, P_n) &= \frac{b-a}{n} \sum_{j=1}^n f(x_j) - f(x_{j-1}) \\ &= \frac{b-a}{n} (f(x_n) - f(x_1)) && \text{[Intermediate terms cancel]} \\ &= \frac{b-a}{n} (f(b) - f(a)). \end{aligned}$$

When  $n \rightarrow \infty$ ,  $U(f, P_n) - L(f, P_n) = \frac{b-a}{n} (f(b) - f(a)) \rightarrow 0$ . Therefore,

$$U(f, P_n) - L(f, P_n) < \varepsilon \quad \text{for large } n \implies f \text{ is integrable.}$$

Q.E.D. ■

**Remark 4.12** To prove a function  $f$  is integrable, it is sufficient to show that  $\forall \varepsilon > 0, \exists$  partition  $P$  s.t.

$$U(f, P) - L(f, P) < \varepsilon.$$

**Theorem 4.5.6 Rules of Integration**

- $k \int_a^b f(x) \, dx = \int_a^b k f(x) \, dx$ ,  $k$  is a constant.
- $\int_a^b (f(x) \pm g(x)) \, dx = \int_a^b f(x) \, dx \pm \int_a^b g(x) \, dx$
- $\int_a^c f(x) \, dx = \int_a^b f(x) \, dx + \int_b^c f(x) \, dx$ , for  $a \leq b \leq c$ .
- If  $f \leq g$ , then  $\int_a^b f(x) \, dx \leq \int_a^b g(x) \, dx$ .

In particular,  $-|f| \leq f \leq |f|$ , so

$$-\int_a^b |f| \leq \int_a^b f \leq \int_a^b |f|.$$

That is,

$$\left| \int_a^b f \right| \leq \int_a^b |f|.$$

**Definition 4.5.7 (Antiderivative).** Let  $f(x) : [a, b] \rightarrow \mathbb{R}$ . An *antiderivative* of  $f$  is a continuous function  $F(x) : [a, b] \rightarrow \mathbb{R}$  s.t.  $F'(x) = f(x) \quad \forall x \in (a, b)$ .

**Remark 4.13 (Antiderivative is not Unique)** Suppose  $F(x)$  is an antiderivative of  $f(x)$ . If  $G$  is another antiderivative, then

$$\frac{d}{dx}[G(x) - F(x)] = G'(x) - F'(x) = f(x) - f(x) = 0 \quad \forall x \in (a, b).$$

So, by MVT,  $G(x) - F(x) = C$ , where  $C$  is some constant, or

$$G(x) = F(x) + C.$$

**Theorem 4.5.8 Fundamental Theorem of Calculus (FTC)**

Let  $f(x) : [a, b] \rightarrow \mathbb{R}$  be continuous. Then,  $f$  has an antiderivative  $F$ , and

$$\int_a^b f(x) \, dx = F(b) - F(a) \left[ = F(x) \Big|_a^b \right].$$

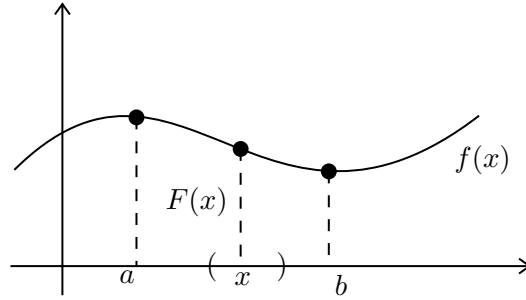
**Proof 3.** Define  $F(x)$  by

$$F(x) = \int_a^x f(t) \, dt$$

for  $x \in [a, b]$ .

**Claim**  $F(x)$  is an antiderivative of  $f(x)$ .

*Proof.*



Fix  $x \in (a, b)$ . Let  $h > 0$  s.t.  $(x - h, x + h) \subset (a, b)$ . Then,

$$\begin{aligned} \frac{F(x+h) - F(x)}{h} &= \frac{1}{h} \left( \int_a^{x+h} f(t) dt - \int_a^x f(t) dt \right) \\ &= \frac{1}{h} \left( \cancel{\int_a^x f(t) dt} + \int_x^{x+h} f(t) dt - \cancel{\int_a^x f(t) dt} \right) = \frac{1}{h} \int_x^{x+h} f(t) dt. \end{aligned}$$

Note that

$$f(x) = \frac{1}{h} \int_x^{x+h} \underbrace{f(x)}_{\text{constant w.r.t. } t} dt$$

So,

$$\begin{aligned} \frac{F(x+h) - F(x)}{h} - f(x) &= \frac{1}{h} \int_x^{x+h} f(t) dt - \frac{1}{h} \int_x^{x+h} f(x) dt \\ &= \frac{1}{h} \int_x^{x+h} f(t) - f(x) dt \end{aligned}$$

Given  $\varepsilon > 0$ ,  $f$  is continuous at  $x$ . So,  $\exists \delta > 0$  s.t.

$$|t - x| < \delta \implies |f(t) - f(x)| < \varepsilon.$$

Then, when  $|h| < \delta$ , we have

$$\begin{aligned} \left| \frac{F(x+h) - F(x)}{h} \right| &\leq \left| \frac{1}{h} \int_x^{x+h} f(t) - f(x) dt \right| \\ &\leq \frac{1}{|h|} \int_x^{x+h} |f(t) - f(x)| dt \\ &< \frac{1}{|h|} \int_x^{x+h} \varepsilon dt \\ &= \frac{1}{|h|} \cdot \varepsilon \cdot |h| \\ &= \varepsilon. \end{aligned}$$

So,

$$\lim_{h \rightarrow 0} \frac{F(x+h) - F(x)}{h} = f(x) \quad \text{i.e., } F'(x) = f(x).$$

Furthermore, one can show that  $F(x)$  is continuous on  $[a, b]$ . [As  $F(x)$  is differentiable on  $(a, b)$ , it is continuous on  $(a, b)$ . We only need to check for the endpoints.]  $\square$

Finally, by definition,

$$F(b) = \int_a^b f(t) \, dt \quad \text{and} \quad F(a) = \int_a^b f(t) \, dt = 0.$$

So,

$$\int_a^b f(t) \, dt = F(b) - F(a).$$

Q.E.D.  $\blacksquare$

**Remark 4.14** In FTC, the continuity assumption of  $f(x)$  cannot be removed. More specifically, it cannot be replaced by integrability. For example,

$$f(x) = \begin{cases} 0, & 0 \leq x \leq 1 \\ 1, & 1 < x \leq 2. \end{cases}$$

$f$  is integrable, and its antiderivative

$$F(x) = \int_0^x f(t) \, dt \quad \text{is well-defined.}$$

However,  $F'(x) = f(x)$  for  $1 < x \leq 2$ . When  $x = 1$ ,  $F'(x)$  does not even exist.



## 5 Uniform Convergence

### 5.1 Definition of Convergence

**Definition 5.1.1 (Pointwise Convergence).** Given a sequence of functions  $f_k(x) : A \subset M \rightarrow N$  for  $k = 1, 2, \dots$ . We say  $f_k(x) \rightarrow f(x)$  *converges pointwise* on  $A$  if  $\forall x \in A$ , the sequence of points  $\{f_k(x)\}$  converges to  $f(x)$ . That is,  $\forall x, \forall \varepsilon > 0, \exists K$  s.t.  $k \geq K \implies \rho(f_k(x), f(x)) < \varepsilon$ .

**Definition 5.1.2 (Uniform Convergence).**  $f_k(x) \rightarrow f(x)$  *converges uniformly* on  $A$  if  $\forall \varepsilon > 0, \exists K$  s.t.  $k \geq K \implies \rho(f_k(x), f(x)) < \varepsilon \quad \forall x \in A$ . We write  $f_k \rightarrow f$  UC on  $A$ .

**Remark 5.1** For pointwise convergence, the choice of  $K$  depends both on  $\varepsilon$  and the point  $x$ . However, for uniform convergence,  $K$  only depends on  $\varepsilon$  but not specific point  $x$ .

**Definition 5.1.3 (Convergence of Series).** Assume  $N$  is a normed space. Suppose  $g_k : A \subset M \rightarrow N$ . Then,  $\sum_{k=1}^{\infty} g_k(x)$  converges to  $g(x)$  *pointwise* or *uniformly*. Using sequence of partial sums, we have

$$f_n(x) = \sum_{k=1}^n g_k(x).$$

**Remark 5.2** UC is stronger:  $UC \implies$  pointwise convergence.

However, pointwise convergence  $\not\Rightarrow$  UC in general.

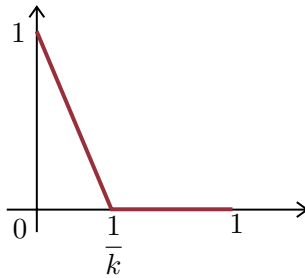
#### Example 5.1.4

Consider  $A = [0, 1]$  and

$$f_k(x) = \begin{cases} 0 & \text{if } \frac{1}{k} \leq x \leq 1 \\ 1 & \text{if } 0 \leq x \leq \frac{1}{k} \end{cases}$$

Note that  $f_k(x) \rightarrow f(x)$  pointwise, where

$$f(x) = \begin{cases} 0 & x > 0 \\ 1 & x = 0. \end{cases}$$



However, this convergence is not uniform:  $\exists \varepsilon_0 > 0$  s.t.  $\forall K, \exists k \geq K$  s.t.  $\rho(f_k(x), f(x)) > \varepsilon_0$  for some  $x \in A$ . For example, take  $\varepsilon_0 = \frac{1}{2}$  and  $0 < x < \frac{1}{k}$ .

**Theorem 5.1.5 Continuity of Uniform Limit**

Let  $f_k : A \subset M \rightarrow N$  be a sequence of continuous functions and  $f_k \rightarrow f$  uniformly converges on  $A$ . Then,  $f$  is also continuous.

**Proof 1.** Fix  $x_0 \in A$ . Given  $\varepsilon > 0$ . By UC,  $\exists K$  s.t.  $\rho(f_K(x), f(x)) < \frac{\varepsilon}{3} \quad \forall x \in A$ . Since  $f_K$  is continuous,  $\exists \delta > 0$  s.t.

$$x \in A, d(x, x_0) < \delta \implies \rho(f_K(x), f(x_0)) < \frac{\varepsilon}{3}.$$

Therefore, by triangle inequality, we have

$$\begin{aligned} \rho(f(x), f(x_0)) &\leq \rho(f(x), f_K(x)) + \rho(f_K(x), f_K(x_0)) + \rho(f_K(x_0), f(x_0)) \\ &< \frac{\varepsilon}{3} + \frac{\varepsilon}{3} + \frac{\varepsilon}{3} \\ &= \varepsilon. \end{aligned}$$

So,  $f$  is continuous at  $x_0$ .

Q.E.D. ■

**Remark 5.3** This result can be used to show that a convergence is not uniform.

**Example 5.1.6**

- $f_n(x) = \frac{x^n}{1+x^n}$ , with  $A = [0, 2]$ .

– Find pointwise limit

$$f_n(x) \rightarrow f(x) = \begin{cases} 0, & 0 \leq x \leq 1 \\ \frac{1}{2}, & x = 0 \\ 1, & 1 < x \leq 2. \end{cases}$$

– Determine uniform convergence:

The convergence is not uniform because  $f$  is not continuous.

- Geometric Series: *Counterexample to the converse of Theorem 5.1.5*

$$\sum_{k=0}^{\infty} x^k \quad \text{with } A = (-1, 1).$$

– Converge pointwise to  $g(x) = \frac{1}{1-x}$ .

Find partial sum:

$$S_n(x) = \sum_{k=0}^n x^k = \frac{1-x^{n+1}}{1-x}.$$

Since  $x \in (-1, 1)$ , as  $n \rightarrow \infty$ ,  $x^{n+1} \rightarrow 0$ . So,

$$S_n(x) = \frac{1 - x^{n+1}}{1 - x} \xrightarrow{n \rightarrow \infty} \frac{1}{1 - x} \quad \text{for } x \in (-1, 1).$$

- Uniform convergence on subinterval  $[-a, a]$  for any  $0 < a < 1$ .

Estimate the error term:

$$|S_n(x) - g(x)| = \frac{|x|^{n+1}}{|1 - x|}.$$

When  $x \rightarrow 1$ ,  $|S_n(x) - g(x)| \rightarrow \infty$  as  $|1 - x| \rightarrow 0$ . However, if we restrict  $x \in [-a, a]$  for some  $0 < a < 1$ , then  $|1 - x| \geq 1 - a$ , and we have

$$\forall \varepsilon > 0, \quad \exists N \text{ s.t. } n \geq N \implies \frac{a^{n+1}}{1 - a} < \varepsilon.$$

$$\implies |S_n(x) - g(x)| \leq \frac{a^{n+1}}{1 - a} < \varepsilon \quad \forall x \in [-a, a].$$

- Convergence is not uniform on  $(-1, 1)$ .

Observe that for any fixed  $N$ , we have  $\frac{|x|^{N+1}}{|1 - x|} \xrightarrow{x \rightarrow 1^-} \infty$ . Therefore,

$$\exists x_0 < 1 \text{ s.t. } \frac{|x_0|^{N+1}}{|1 - x_0|} = |S_N(x_0) - g(x_0)| \geq 1 = \varepsilon_0.$$

**Definition 5.1.7 (Uniformly Cauchy Sequence).** A sequence of functions  $f_k : A \subset M \rightarrow N$  is *uniformly Cauchy sequence* if  $\forall \varepsilon > 0, \exists L > 0 \text{ s.t. } j, k \geq L \implies \rho(f_k(x), f_j(x)) < \varepsilon \quad \forall x \in A$ .

### Theorem 5.1.8 Cauchy Criterion

Let  $(N, \rho)$  be a *complete* metric space and  $f_k : A \subset M \rightarrow N$  be a sequence of functions. Then,  $f_k$  converges uniformly on  $A \iff \forall \varepsilon > 0, \exists L > 0 \text{ s.t.}$

$$j, k \geq L \implies \rho(f_k(x), f_j(x)) < \varepsilon \quad \forall x \in A.$$

**Proof2.** ( $\Rightarrow$ ) Assume  $f_k \rightarrow f$  uniformly. [WTS:  $f_k$  is uniformly Cauchy.]

$$\rho(f_k(x), f_j(x)) \leq \rho(f_k(x), f(x)) + \rho(f(x), f_j(x)). \quad \square$$

( $\Leftarrow$ ) Assume  $\{f_k\}$  is uniformly Cauchy.

- Find the limit function (pointwise)

For each fixed  $x \in A$ , the sequence of points  $\{f_k(x)\}$  is Cauchy in  $N$ . By completeness of  $N$ ,  $f_k(x)$  converges to some point in  $N$ . Denoted by  $f(x)$ .

- **Show  $f_k(x) \rightarrow f(x)$  UC**

Given  $\varepsilon > 0$ ,  $\exists L_1$  s.t.  $j, k \geq L_1 \implies \rho(f_k(x), f_j(x)) < \frac{\varepsilon}{2} \quad \forall x \in A$ . Furthermore, as  $f_k(x) \rightarrow f(x)$  pointwise, for any  $x \in A$ ,  $\exists L_x \geq L_1$  s.t.  $j \geq L_x \implies \rho(f_j(x), f(x)) < \frac{\varepsilon}{2}$ .

Now, let  $K = L_1$ . Then, when  $k \geq K$  we have

$$\begin{aligned} \rho(f_k(x), f(x)) &\leq \rho(f_k(x), f_{L_x}(x)) + \rho(f_{L_x}(x), f(x)) \\ &< \frac{\varepsilon}{2} + \frac{\varepsilon}{2} \\ &= \varepsilon \quad \forall x \in A. \end{aligned}$$

Just pick  $j = L_x$ , we have different intermediate term for different  $x$ 's.

Q.E.D. ■

**Corollary 5.1.9 Weierstrass  $M$  Test:** Let  $N$  be a complete normed space and  $g_k : A \rightarrow N$  be a sequence of functions s.t.  $\exists$  constants  $M_k$  with

- $\|g_k(x)\| \leq M_k$  for all  $x \in A$ , and
- $\sum_{k=1}^{\infty} M_k$  converges.

Then, the series  $\sum_{k=1}^{\infty} g_k(x)$  converge uniformly.

**Proof 3.** The sequence of partial sums  $\{f_n(x)\}$  is uniformly Cauchy.

$$f_n(x) = \sum_{k=1}^n g_k(x).$$

Then, apply Cauchy criterion.

Q.E.D. ■

#### Example 5.1.10

- $\sum_{n=1}^{\infty} \frac{(\sin nx)^2}{n^2}, \quad A = \mathbb{R}.$

Set  $g_n(x) = \frac{(\sin nx)^2}{n^2}$ . Then,  $|g_n(x)| \leq \frac{1}{n^2}$ .

As  $\sum_{n=1}^{\infty} \frac{1}{n^2}$  converges, by  $M$  test,  $\sum_{n=1}^{\infty} \frac{(\sin nx)^2}{n^2}$  converges uniformly.

- $\sum_{n=0}^{\infty} \left(\frac{x^n}{n!}\right)^2 \rightarrow f(x)$  on  $\mathbb{R}$  pointwise

If we limit  $A = [-a, a]$ , then  $\sum_{n=0}^{\infty} \left(\frac{x^n}{n!}\right)^2$  uniformly converges.

## 5.2 Integration and Differentiation of Series

### Theorem 5.2.1

Suppose  $f_n : [a, b] \rightarrow \mathbb{R}$  and integrable and  $f_n \rightarrow f$  uniformly on  $[a, b]$ . Then,  $f$  is integrable, and

$$\lim_{n \rightarrow \infty} \int_a^b f_n(x) \, dx = \int_a^b \lim_{n \rightarrow \infty} f_n(x) \, dx = \int_a^b f(x) \, dx.$$

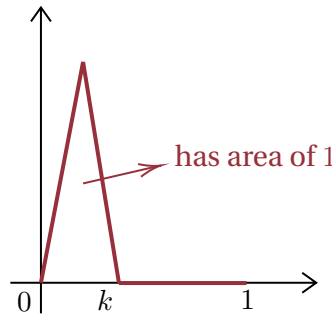
**Proof 1.** Assume  $f$  is integrable. Then,

$$\begin{aligned} \left| \int_a^b f_n(x) \, dx - \int_a^b f(x) \, dx \right| &\leq \int_a^b \underbrace{|f_n(x) - f(x)|}_{< \varepsilon \quad \forall x, \text{ by UC}} \, dx \\ &< \int_a^b \varepsilon \, dx = \varepsilon(b - a). \end{aligned}$$

Q.E.D. ■

**Remark 5.4** One cannot replace uniform convergence by pointwise convergence.

### Example 5.2.2



Define  $f_k(x) : [0, 1] \rightarrow \mathbb{R}$  s.t.

$$\int_0^1 f_k(x) \, dx = 1.$$

Observe that  $f_k(x) \xrightarrow{\text{pointwise}} f(x) \equiv 0 \quad \forall x \in [0, 1]$ . So,

$$\int_0^1 f_k(x) \, dx \not\rightarrow \int_0^1 f(x) \, dx$$

**Remark 5.5** The same result is not true for differentiation. One cannot simply replace integrable with differentiable. For example, consider

$$f_n(x) = \frac{x^{n+1}}{n+1} \quad \text{on } [0, 1] \implies f'_n(x) = x^n.$$

We have  $f_n(x) \xrightarrow{UC} f(x) \equiv 0$ . However,

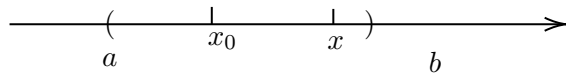
$$\lim_{n \rightarrow \infty} f'_n(x) \neq \lim_{n \rightarrow \infty} f'(x).$$

### Theorem 5.2.3

Let  $f_n : (a, b) \rightarrow \mathbb{R}$  be differentiable, converging pointwise to  $f(x) : (a, b) \rightarrow \mathbb{R}$ . If  $f'_n(x)$  are continuous and converges uniformly to a function  $g$ , then  $f'(x) = g(x)$ . i.e.,

$$\lim_{n \rightarrow \infty} \frac{d}{dx}(f_n(x)) = \frac{d}{dx} \left( \lim_{n \rightarrow \infty} f_n(x) \right) = \frac{d}{dx} f(x) = g(x).$$

**Proof2.**



Use Fundamental Theorem of Calculus,

$$\begin{aligned} f_n(x) &= f_n(x_0) + f_n(x) - f_n(x_0) \\ &= f_n(x_0) + \int_{x_0}^x f'_n(t) dt. \end{aligned}$$

When  $n \rightarrow \infty$ , for fixed  $x \in A$ ,

$$f_n(x) \rightarrow f(x), \quad f_n(x_0) \rightarrow f(x_0), \quad \int_{x_0}^x f'_n(t) dt \rightarrow \int_{x_0}^x g(t) dt.$$

So,

$$\begin{aligned} f(x) &= f(x_0) + \int_{x_0}^x g(t) dt \\ \frac{d}{dx}(f(x)) &= \frac{d}{dX}(f(x_0)) + \frac{d}{dx} \int_{x_0}^x g(t) dt \\ \lim_{n \rightarrow \infty} f'_n(x) &= f'(x) = 0 + g(x) = g(x). \end{aligned}$$

Q.E.D. ■

### Example 5.2.4 One cannot replace UC with pointwise convergence

$$f_n = \frac{nx^2}{1 + nx^2}, \quad -1 \leq x \leq 1 \implies f'_n(x) \xrightarrow{\text{pointwise}} g(x)$$

However,  $f'_n(x) \neq g(x)$ .

## 5.3 The Space of Continuous Functions

**Notation 5.1.** Let  $A \subset M$  be a metric space and  $N$  is a normal vector space. Then

- $\mathcal{C} = \mathcal{C}(A, N) = \{f \mid f: A \rightarrow N \text{ continuous}\}$ : the collection of all continuous functions  $f: A \rightarrow N$
- $\mathcal{C}_b = \mathcal{C}(A, N) = \{f \in \mathcal{C} \mid f \text{ is bounded}\}$ : the collection of all bounded continuous functions  
( $\exists M$  s.t.  $\|f(x)\|_N \leq M \quad \forall x \in A$ )

**Example 5.3.2**

$A = [0, 1] \subset \mathbb{R}, N = \mathbb{R}$ . Then,

$$\mathcal{C}_b = \mathcal{C}, \quad \text{the set of all continuous functions on } [0, 1].$$

**Remark 5.6**

- $\mathcal{C}_b$  and  $\mathcal{C}$  are vector spaces;
- Goal: Study  $\mathcal{C}_b$  as a normed vector spaces as  $\mathbb{R}^n$ .

**Definition 5.3.3 (Norm on  $\mathcal{C}_b$ ).** Given  $f \in \mathcal{C}_b$ . Define  $\|f\|$  as follows:

$$\|f\| = \sup \{\|f(x)\|_N \mid x \in A\}.$$

This is called the *maximum absolute value norm*.

**Theorem 5.3.4**

$\|\cdot\|$  defined in Definition 5.3.3 is a norm in  $\mathcal{C}_b$ . i.e.,

- Positive definiteness:  $\|f\| \geq 0$  and  $\|f\| = 0 \iff f = 0$ ;
- Scalar multiplicity:  $\|\alpha f\| = |\alpha| \|f\| \quad \forall \alpha \in \mathbb{R}$
- Triangle inequality:  $\|f + g\| \leq \|f\| + \|g\|$

**Proof 1.** (of ③) By definition,  $\|f + g\| = \sup \{\|f(x) + g(x)\|_N \mid x \in A\}$ . [WTS:  $\|f\| + \|g\|$  is an upper bound.] Note that

$$\begin{aligned} \|f(x) + g(x)\|_N &\leq \|f(x)\|_N + \|g(x)\|_N && \text{[triangle inequality in } N\text{]} \\ &\leq \|f\| + \|g\| && \text{[definition]} \end{aligned}$$

So,  $\|f + g\| \leq \|f\| + \|g\|$ .

Q.E.D. ■

**Definition 5.3.5 (Convergence in  $\mathcal{C}_b$ ).**  $f_k \rightarrow f$  in  $\mathcal{C}_b$  means that  $\|f_k - f\| \rightarrow 0$  as  $k \rightarrow \infty$ .

**Theorem 5.3.6**

$f_k \rightarrow f$  in  $\mathcal{C}_b$  (convergence in norm as vectors)  $\iff f_k \rightarrow f$  uniformly on  $A$  (convergence in function)

**Proof2.** ( $\Rightarrow$ ): Assume  $\|f_k - f\| \rightarrow 0$ . Then,  $\forall \varepsilon > 0, \exists K$  s.t.  $k \geq K \implies \|f_k - f\| \leq \varepsilon$ . Thus,  $\forall x \in A$ , by definition of norm, for  $k \geq K$ ,

$$\|f_k(x) - f(x)\|_N \leq \|f_k - f\| < \varepsilon.$$

So,  $f_k(x) \rightarrow f(x)$  uniformly on  $A$ .  $\square$

( $\Leftarrow$ ): Assume  $f_k(x) \rightarrow f(x)$  uniformly on  $A$ . Then,  $\forall \varepsilon > 0, \exists K$  s.t.  $k \geq K \implies \|f_k(x) - f(x)\|_N < \varepsilon$ . Then,  $\varepsilon$  is an upper bound. Note that

$$\|f_k - f\| = \sup \{\|f_k(x) - f(x)\|_N \mid x \in A\}$$

is a least upper bound. So,

$$\|f_k - f\| = \sup \{\|f_k(x) - f(x)\|_N \mid x \in A\} < \varepsilon$$

So,  $\|f_k - f\| \rightarrow 0$  as  $k \rightarrow \infty$ .

Q.E.D.  $\blacksquare$

### Theorem 5.3.7 Completeness of $\mathcal{C}_b$

If  $N$  is complete, so is  $\mathcal{C}_b(A, N)$ .

**Proof3.** Let  $\{f_k\}$  be a Cauchy sequence in  $\mathcal{C}_b$ . Then,  $\forall \varepsilon > 0, \exists K$  s.t.  $j, k \geq K \implies \|f_j - f_k\| < \varepsilon$ . By definition, we have

$$\|f_j(x) - f_k(x)\|_N \leq \|f_j - f_k\| < \varepsilon \quad \forall x \in A.$$

So,  $\{f_k(x)\}$  is a uniform Cauchy sequence on  $A$ . By Cauchy criterion,

$$f_k(x) \rightarrow f(x) \quad \text{uniformly on } A.$$

$f$  is also continuous since UC preserves continuity. By Theorem 5.3.6, we have  $f_k \rightarrow f$  in  $\mathcal{C}_b$ . So,  $\mathcal{C}_b$  is complete.

Q.E.D.  $\blacksquare$

**Remark 5.7 (Comparison Between  $\mathcal{C}_b$  and  $\mathbb{R}^n$ )** Let  $A \subset M$  be compact and  $N = \mathbb{R}^n$ .

Properties	$\mathbb{R}^n$	$\mathcal{C}_b(A, N = \mathbb{R}^n)$
Normed Space	✓	✓
Completeness	✓	✓
Finite Dimension	✓	✗
Compact Subset	<u>Heine-Borel:</u> $B \subset \mathbb{R}^n$ is compact $\iff B$ is closed and bounded	<u>Arzela-Ascoli:</u> $A \subset M$ compact. Then, $\mathcal{B} \subset \mathcal{C}_b$ is compact $\iff \mathcal{B}$ is closed, bounded, and equicontinuous in $A$



**Definition 5.3.8 (Equicontinuous).** A family of function  $\mathcal{B}$  is equicontinuous at a point  $x \in A$  if  $\forall \varepsilon > 0, \exists \delta > 0$  s.t.  $y \in D(x, \delta) \cap A \implies \|f(x) - f(y)\|_N < \varepsilon \quad \forall f \in \mathcal{B}$ .

**Remark 5.8**  $\delta$  is independent of  $f \in \mathcal{B}$ .

### Example 5.3.9

- $\mathcal{B} = \{f \in \mathcal{C}_b(\mathbb{R}, \mathbb{R}) \mid f(x) > 0 \quad \forall x \in \mathbb{R}\}.$

– Is  $\mathcal{B}$  open? No.

Suppose  $f \rightarrow 0$  as  $x \in \infty$ . Then, no matter how small we take the  $\delta$ , some part of  $D(f, \delta)$  will not be contained in  $\mathcal{B}$ .

– What is  $\text{cl}(\mathcal{B})$ ?

$$\text{cl}(\mathcal{B}) = \{f \in \mathcal{C}_b(\mathbb{R}, \mathbb{R}) \mid f(x) \geq 0 \quad \forall x \in \mathbb{R}\}.$$

– What is  $\text{int}(\mathcal{B})$ ?

$$\text{int}(\mathcal{B}) = \{f \in \mathcal{C}_b(\mathbb{R}, \mathbb{R}) \mid \inf(f(x)) > 0 \quad \forall x \in \mathbb{R}\}.$$

Think of  $\inf(f(x)) > 0$  in this way: we need a buffer zone.

- $\mathcal{B} = \{f \in \mathcal{C}_b([0, 1], \mathbb{R}) \mid f(x) > 0 \quad \forall x \in [0, 1]\}.$

## 5.4 The Contraction Mapping Principle (CMP)

### Theorem 5.4.1 CMP

Let  $(M, d)$  be a complete metric space, and  $\Phi : M \rightarrow M$  be a map. Suppose  $\exists$  constant  $k$  s.t.  $0 < k < 1$  s.t.

$$d(\Phi(x), \Phi(y)) \leq k \cdot d(x, y) \quad \forall x, y \in M.$$

Then,

- $\Phi$  has a unique fixed point in  $M$ . That is,  $\exists ! x^* \in M$  s.t.  $\Phi(x^*) = x^*$ .
- The fixed point can be constructed (or approximated) as follows:

Fix any point  $x_0 \in M$ . Let  $x_1 = \Phi(x_0), x_2 = \Phi(x_1), \dots, x_{n+1} = \Phi(x_n), \dots$ . Then,

$$\lim_{n \rightarrow \infty} x_n = x^*.$$

**Remark 5.9**  $\Phi$  is continuous. Further,  $\Phi$  is Lipschitz  $\implies \Phi$  is uniform continuous.

**Proof 1.** Fix  $x_0 \in M$ . Let  $x_{n+1} = \Phi(x_n)$  for  $n = 0, 1, 2, \dots$ .

**Claim**  $\{x_n\}$  is Cauchy.

Note that  $\forall n \geq 1$ ,

$$\begin{aligned} d(x_n, x_{n+1}) &= d(\Phi(x_{n-1}), \Phi(x_n)) \leq kd(x_{n-1}, x_n) \\ &\leq k^2 d(x_{n-1}, x_{n-1}) \\ &\vdots \\ &\leq k^n d(x_0, x_1). \end{aligned}$$

Thus,  $\forall p \geq 1$ ,

$$\begin{aligned} d(x_n, x_{n+p}) &\leq d(x_n, x_{n+1}) + d(x_{n+1}, x_{n+2}) + \cdots + d(x_{n+p-1}, x_{n+p}) \\ &\leq k^n d(x_0, x_1) + k^{n+1} d(x_0, x_1) + \cdots + k^{n+p-1} d(x_0, x_1) \\ &= \underbrace{(k^n + k^{n+1} + \cdots + k^{n+p-1})}_{\text{geometric series}} d(x_0, x_1) \rightarrow 0 \quad \text{as } n \rightarrow \infty. \end{aligned}$$

As the geometric series converges,  $\{x_n\}$  is Cauchy.

Since  $M$  is complete,  $x_n \rightarrow x^* \in M$ .

**Claim**  $x^*$  is a fixed point.

Since  $\Phi$  is continuous,

$$\lim_{n \rightarrow \infty} \Phi(x_n) = \Phi\left(\lim_{n \rightarrow \infty} x_n\right) = \Phi(x^*).$$

Meanwhile,  $\Phi(x_n) = x_{n+1}$ , so

$$\lim_{n \rightarrow \infty} \Phi(x_n) = \lim_{n \rightarrow \infty} x_{n+1} = x^*.$$

Hence,  $x^* = \Phi(x^*)$ , implying  $x^*$  is a fixed point.

**Claim** The fixed point is unique.

Let  $y^* \in M$  be another fixed point. One can show

$$\begin{aligned} d(x^*, y^*) &\leq d(\Phi(x^*), \Phi(y^*)) && [x^*, y^* \text{ are fixed points}] \\ &\leq kd(x^*, y^*) && [\Phi \text{ is a contraction mapping}] \end{aligned}$$

$$\implies d(x^*, y^*) = 0.$$

Q.E.D. ■

#### Example 5.4.2 Application in ODE

Consider the following initial value problem (IVP):

$$\frac{dx}{dt} = f(t, x) \quad x(t_0) = x_0 \quad (\text{IVP})$$

- Basic Assumptions:

1.  $f(t, x)$  is continuous in a neighborhood  $U$  of  $(t_0, x_0) \in \mathbb{R}^2$
2.  $f(t, x)$  is Lipschitz in  $x$ :  $\exists$  constant  $K$  s.t.

$$|f(t, x_1) - f(t, x_2)| \leq K|x_1 - x_2| \quad \forall (t_1, x_1), (t_1, x_2) \in U$$

- Apply CMP:

**Theorem 5.4.3**

If  $f(t, x)$  is continuous in  $U$  and Lipschitz in  $x$ , then (IVP) has a unique solution  $x = \varphi(t)$  in the neighborhood of  $t_0$ :  $(t_0 - \delta, t_0 + \delta)$ . i.e.,

$$\varphi'(t) = f(t, \varphi(t)), \quad \varphi(t_0) = x_0.$$

- Solving (IVP) is equivalent to finding a function  $\varphi(t)$  s.t.

$$\varphi'(t) = f(t, \varphi(t)).$$

Or, by integration:

$$\varphi'(t) = x_0 + \int_{t_0}^t f(s, \varphi(s)) \, ds \quad [x_0 \text{ comes from plugging in the initial condition}]$$

This is just a fixed point for the following map (an integral operator):

$$\Phi : g(t) \mapsto \Phi(g) = x_0 + \int_{t_0}^t f(s, g(s)) \, ds$$

**Theorem 5.4.4**

We need to construct an appropriate metric space  $M \subset \mathcal{C}_b$  s.t.  $\Phi : M \rightarrow M$  is a contraction mapping.

---

**Algorithm 1:** Iterative Method to Approximate the Solution to (IVP)

---

```

1 begin
2    $\varphi_0 \equiv x_0$ ;
3   for  $n = 0, 1, 2, \dots$  do
4      $\varphi_{n+1}(t) = \Phi(\varphi_n(t)) = x_0 + \int_{t_0}^t f(s, \varphi_n(s)) \, ds$ ;
```

---

**Example 5.4.5**

Consider the IVP:  $f(t, x) = tx^2 + x^3$ ,  $x(0) = 1$ .

Let  $\varphi_0(t) = 1$ . Then,

$$\begin{aligned}
 \varphi_1(t) &= 1 + \int_0^t s\varphi_0(s)^2 + \varphi_0(s)^3 \, ds \\
 &= 1 + \int_0^t s + 1 \, ds \\
 &= 1 + \left[ \frac{1}{2}s^2 + s \right]_0^t \\
 &= 1 + \frac{1}{2}t^2 + t \\
 \varphi_2(t) &= 1 + \int_0^t s\varphi_1(s)^2 + \varphi_1(s)^3 \, ds \\
 &= 1 + \int_0^t s \left( 1 + \frac{1}{2}s^2 + s \right)^2 + \left( 1 + \frac{1}{2}s^2 + s \right)^3 \, ds \\
 &\vdots
 \end{aligned}$$

## 6 Differential Mappings

### 6.1 Definition and Matrix Representation of a Differential

**Definition 6.1.1 (Linear Transformation).** A function  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is called a *linear transformation* if  $\forall x, y \in \mathbb{R}^n$  and  $\lambda \in \mathbb{R}$ , we have

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

These two properties can be combined and written equivalently as  $T(ax + by) = aT(x) + bT(y) \quad \forall x, y \in \mathbb{R}^n$  and  $\forall a, b \in \mathbb{R}$ .

#### 6.1.2 Matrix Representation of $T$ .

**Observation:** Given  $m \times n$  matrix  $A$ , define function  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  by  $T(x) = A \cdot x$ . Then,  $T$  is a linear transformation.

**Proof 1.**

$$T(ax + by) = A(ax + by) = A(ax) + A(by) = aAx + bAy = aT(x) + bT(y).$$

Q.E.D. ■

#### Example 6.1.3

Suppose  $A = \begin{bmatrix} 2 & 3 \\ 1 & -1 \\ 0 & 4 \end{bmatrix}$ . Then,

$$T(x) = A \cdot x = \begin{bmatrix} 2 & 3 \\ 1 & -1 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2x_1 + 3x_2 \\ x_1 - x_2 \\ 4x_2 \end{bmatrix} \in \mathbb{R}^3.$$

#### Theorem 6.1.4 Fact

Every linear transformation  $T$  is determined by a matrix in such a way as above (via matrix multiplication).

**Proof 2.** Given  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  linear, we need to find a matrix  $A$  ( $m \times n$ ) such that

$$T(x) = A \cdot x \quad \forall x \in \mathbb{R}^n.$$

To construct  $A$ , consider the standard basis for  $\mathbb{R}^n : \{e_1, e_2, \dots, e_n\}$  and for  $\mathbb{R}^m : \{e'_1, e'_2, \dots, e'_m\}$ . Then,

$$T(e_j) = \sum_{i=1}^m a_{ij} e'_i, \quad \forall j = 1, 2, \dots, n.$$

Let  $A = (a_{ij})_{m \times n}$ .

**Claim**  $T(x) = Ax \quad \forall x \in \mathbb{R}^n$ .

In fact, let  $x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n$ . Then, we can rewrite  $x$  as a linear combination of standard basis:

$$x = \sum_{j=1}^n x_j e_j.$$

So,

$$T(x) = x_1 T(e_1) + x_2 T(e_2) + \dots + x_n T(e_n) = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = Ax. \quad [T \text{ is Linear}]$$

Q.E.D. ■

**Remark 6.1** The collection of  $\{\text{linear transformation } T : \mathbb{R}^n \rightarrow \mathbb{R}^m\}$  forms a 1-to-1 correspondence with the collection of  $\{m \times n \text{ matrices } A\}$ .

#### Theorem 6.1.5 Continuity of $T$

If  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is linear, then it is Lipschitz, and hence continuous.

**Proof 3.** Recall the definition of Lipschitz:  $|f(x) - f(y)| \leq L \cdot |x - y|$ .

Since  $T(x) - T(y) = T(x - y)$ , we only need to show that

$$\|T(x)\| \leq M \cdot \|x\| \text{ for some } M \in \mathbb{R}.$$

Let  $x = \sum x_j e_j$ . Then,  $T(x) = \sum x_j T(e_j)$ . So,  $\|T(x)\| \leq \sum |x_j| \cdot \|T(e_j)\|$ .

Recall that  $\|x\| = \sqrt{\sum_j x_j^2}$ . So,  $|x_j| \leq \|x\|$ . Hence,

$$\|T(x)\| \leq \sum_j \|x\| \cdot \|T(e_j)\| = \underbrace{\left( \sum_{j=1}^n \|T(e_j)\| \right)}_{M, \text{ independent of } x} \cdot \|x\| = M \cdot \|x\|$$

Q.E.D. ■

### 6.1.6 Derivative (Differential) as a Linear Transformation.

- Recall one variable case: Let  $f : (a, b) \rightarrow \mathbb{R}$ . Then, we can rewrite  $f'(x_0) = \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0}$  as

$$\lim_{x \rightarrow x_0} \left[ \frac{f(x) - f(x_0) - f'(x_0)(x - x_0)}{x - x_0} \right] = 0.$$

- Definition 6.1.7 (Generalization to Higher Dimensions).** A map  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  is said to be *differentiable* at  $x_0 \in A$  if there is a linear map, denoted by  $\mathbb{D}f(x_0) : \mathbb{R}^n \rightarrow \mathbb{R}^m$  with

$$\lim_{x \rightarrow x_0} \frac{\|f(x) - f(x_0) - \mathbb{D}f(x_0)(x - x_0)\|}{\|x - x_0\|} = 0 \quad (\star)$$

**Remark 6.2** Interpretations of  $(\star)$ :

1. Rewrite  $(\star)$ :  $\forall \varepsilon > 0, \exists \delta > 0$  s.t.  $\forall x \in A$ ,

$$\|x - x_0\| > \delta \implies \|f(x) - f(x_0) - \mathbb{D}f(x_0)(x - x_0)\| < \varepsilon \|x - x_0\|.$$

2.  $f(x) \approx f(x_0) + \underbrace{\mathbb{D}f(x_0) \cdot (x - x_0)}_{\text{linear map}}$  is called the affine map.
3. **Geometric Interpretation:**  $z = f(x) : \mathbb{R}^n \rightarrow \mathbb{R}^1$ . Then,  $z - f(x_0) = \mathbb{D}f(x_0)(x - x_0)$  represents the tangent plane of the surface  $z = f(x)$ .
4. For  $f : \mathbb{R}^1 \rightarrow \mathbb{R}^1$ ,  $\mathbb{D}f(x)$  is the differential, representing a linear transformation, whereas  $f'(x)$  or  $\frac{df}{dx}$  is the derivative, which is just a number.  
For example,  $f(x) = x^2$ . Then,  $f'(x) = 2x$ . However,  $\mathbb{D}f(x)$  is a linear transformation  $\mathbb{R}^1 \rightarrow \mathbb{R}^1$ , defined as

$$\mathbb{D}f(x)(h) = 2xh, \quad \forall h \in \mathbb{R}^1.$$

- Uniqueness of Differential

#### Theorem 6.1.8

Let  $A \subset \mathbb{R}^n$  be open and  $f : A \rightarrow \mathbb{R}^m$  be differentiable at  $x_0 \in A$ . Then, the differential  $\mathbb{D}f(x_0)$  is uniquely determined by  $f$ .

**Proof 4.** Let  $L_1$  and  $L_2$  be two linear transformations such that

$$\lim_{x \rightarrow x_0} \frac{\|f(x) - f(x_0) - L_1(x - x_0)\|}{\|x - x_0\|} = 0 = \lim_{x \rightarrow x_0} \frac{\|f(x) - f(x_0) - L_2(x - x_0)\|}{\|x - x_0\|}.$$

We need to show that  $L_1 = L_2$ . i.e.,  $L_1(h) = L_2(h) \quad \forall h \in \mathbb{R}^n$ .

Fix any unit vector  $e \in \mathbb{R}^n$ . Let  $x = x_0 + te$ , where  $t \in \mathbb{R}$  and  $t \neq 0$  (*This makes sense because  $A$  is open by assumption*). Then,

$$\begin{aligned} \|L_1(e) - L_2(e)\| &= \frac{\|L_1(te) - L_2(te)\|}{|t|} \\ &= \frac{\|L_1(x - x_0) - L_2(x - x_0)\|}{\|x - x_0\|} && [|t| = \|x - x_0\|] \\ &= \frac{\|L_1(x - x_0) - (f(x) - f(x_0)) + (f(x) - f(x_0)) - L_2(x - x_0)\|}{\|x - x_0\|} \\ &\leq \frac{\|L_1(x - x_0) - (f(x) - f(x_0))\| + \|(f(x) - f(x_0)) - L_2(x - x_0)\|}{\|x - x_0\|} \\ &= \frac{\|L_1(x - x_0) - (f(x) - f(x_0))\|}{\|x - x_0\|} + \frac{\|(f(x) - f(x_0)) - L_2(x - x_0)\|}{\|x - x_0\|}. \end{aligned}$$

Note that both parts  $\rightarrow 0$  as  $x \rightarrow x_0$ . So,  $\|L_1(e) - L_2(e)\| = 0$ , and thus  $L_1(e) = L_2(e) \quad \forall$  unit vector  $e$ . Using linear transformation,  $L_1(h) = L_2(h) \quad \forall h \in \mathbb{R}^n$ .

Q.E.D. ■

**Remark 6.3** Theorem 6.1.8 is not true if  $A$  is not open. A trivial example would be when  $A = \{x_0\}$ , the set of just one point. Then, any linear map satisfies the differential definition. That is,

$$\lim_{\substack{x \rightarrow x_0 \\ x \in A}} \frac{\|f(x) - f(x_0) - T(x - x_0)\|}{\|x - x_0\|} = 0 \quad \forall \text{ linear map } T.$$

Or, equivalently,  $\|f(x) - f(x_0) - T(x - x_0)\| < \varepsilon \|x - x_0\|$ .

### 6.1.9 Matrix Representation of the Differential $\mathbb{D}f(x)$ .

**Question:** Given  $f$ , how do we find the linear transformation  $\mathbb{D}f(x)$ ?

**Definition 6.1.10 (Partial Derivative).** Write  $f(x) = (f_1(x_1, \dots, x_n), f_2(x_1, \dots, x_n), \dots, f_m(x_1, \dots, x_n)) \in \mathbb{R}^m$ . Then,

$$\frac{\partial f_j}{\partial x_i} = \lim_{h \rightarrow 0} \frac{f_j(x_1, \dots, x_{i-1}, x_i + h, x_{i+1}, \dots, x_n) - f_j(x_1, \dots, x_i, \dots, x_n)}{h}.$$



**Theorem 6.1.11 Relation Between Differential  $\mathbb{D}f(x)$  and Partial Derivatives**

Suppose  $A \subset \mathbb{R}^n$  is open and  $f : A \rightarrow \mathbb{R}^m$  is differentiable at  $x \in A$ . Then,  $\frac{\partial f_j}{\partial x_i}$  exists and the matrix of the linear map  $\mathbb{D}f(x)$  is given by

$$\begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \frac{\partial f_m}{\partial x_2} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix} \in \mathbb{R}^{m \times n},$$

and we denote this matrix as  $J_f(x)$ , the *Jacobian matrix* of  $f$  at  $x$ .

**Proof 5.** Denote the matrix of  $\mathbb{D}f(x)$  by  $B = (b_{ji})_{m \times n}$ . We need to show  $b_{ji} = \frac{\partial f_j}{\partial x_i}$ .

Recall:  $b_{ji} = j$ -th component of  $\mathbb{D}f(x)(e_i) = \sum_{j=1}^m b_{ji}e'_j$ . Fix  $i, j$  and let  $y = x + he_i$ ,  $h \in \mathbb{R}$ . Then, by definition of differential,

$$\frac{\|f(y) - f(x) - \mathbb{D}f(x)(y - x)\|}{\|y - x\|} \rightarrow 0 \quad \text{as } y - x \rightarrow 0.$$

Taking the  $j$ -th component,

$$\frac{|f_j(x_1, \dots, x_i + h, \dots, x_n) - f_j(x_1, \dots, x_n) - b_{ji} \cdot h|}{|h|} \rightarrow 0 \quad \text{as } h \rightarrow 0.$$

So,

$$\lim_{h \rightarrow 0} \frac{f_j(x_1, \dots, x_i + h, \dots, x_n) - f_j(x_1, \dots, x_n)}{h} = b_{ji}.$$

Hence,

$$\frac{\partial f_j}{\partial x_i} = b_{ji} \quad \forall i, j.$$

So,  $\mathbb{D}f(x)$  is determined by the Jacobian matrix  $J_f(x)$ .

Q.E.D. ■

**Example 6.1.12**

- $f(x, y, z) = (x^4y, xe^z) : \mathbb{R}^3 \rightarrow \mathbb{R}^2$ .

$$J_f(x, y, z) = \begin{bmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} & \frac{\partial f_1}{\partial z} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} & \frac{\partial f_2}{\partial z} \end{bmatrix} = \begin{bmatrix} 4x^3y & x^4 & 0 \\ e^z & 0 & xe^z \end{bmatrix}.$$

- Special Case:  $m = 1$ :  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ . Then,

$$J_f(x) = \begin{bmatrix} \frac{\partial f}{\partial x_1} & \frac{\partial f}{\partial x_2} & \cdots & \frac{\partial f}{\partial x_n} \end{bmatrix} \text{ is a } 1 \times n \text{ matrix.}$$

**Definition 6.1.13 (Gradient).** The *gradient*,  $\text{grad } f$  or  $\nabla f$ , is defined by the following vector:

$$\nabla f = \left( \frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}, \dots, \frac{\partial f}{\partial x_n} \right).$$

Gradient points towards the direction of fastest growth.

- $f(x, y, z) = \frac{x \sin y}{z}$ . Computing  $\mathbb{D}f$  and  $\nabla f$ .

**Solution 6.**

$$\mathbb{D}f(x) = J_f(x) = \begin{bmatrix} \frac{\sin y}{z} & \frac{x \cos y}{z} & -\frac{x \sin y}{z^2} \end{bmatrix}.$$

$$\nabla f(x) = \left( \frac{\sin y}{z}, \frac{x \cos y}{z}, -\frac{x \sin y}{z^2} \right).$$

□

**Remark 6.4 (Relation Between  $\mathbb{D}f(x)$  and  $\nabla f$ )** For any  $h \in \mathbb{R}^n$ , we have

$$\text{matrix multiplication} \leftarrow \mathbb{D}f(x)h = \langle \nabla f, h \rangle \rightarrow \text{inner product/dot product}$$

- Special Case:  $n = 1$ . Consider  $x = c(t) : [a, b] \subset \mathbb{R} \rightarrow \mathbb{R}^m$ . Then,

$$\mathbb{D}x(t) = c'(t) = (c'_1(t), c'_2(t), \dots, c'_m(t))$$

is the tangent vector.

## 6.2 Necessary and Sufficient Conditions for Differentiability

**Definition 6.2.1 (Locally Lipschitz).**  $f$  is *locally Lipschitz* at  $x_0$  if  $\forall x_0 \in A, \exists \delta > 0$  and  $M$  s.t.

$$\|x - x_0\| < \delta \implies \|f(x) - f(x_0)\| < M \cdot \|x - x_0\|.$$

### Theorem 6.2.2 Necessary Condition for Differentiability I

Suppose  $A \subset \mathbb{R}^n$  is open and  $f : A \rightarrow \mathbb{R}^m$  is differentiable. Then,  $f$  is locally Lipschitz.

**Remark 6.5 (Ideas to Prove this Theorem)**

- *Linear map  $\mathbb{D}f(x)$  is Lipschitz;*
- *$f(x)$  can be approximated by  $\mathbb{D}f(x_0)$  locally.*

**Proof 1.** Fix  $x_0 \in A$ . By definition,

$$\lim_{x \rightarrow x_0} \frac{\|f(x) - f(x_0) - \mathbb{D}f(x_0)(x - x_0)\|}{\|x - x_0\|} = 0.$$

For  $\varepsilon = 1$ ,  $\exists \delta > 0$  s.t.

$$\|x - x_0\| < \delta \implies \|f(x) - f(x_0) - \mathbb{D}f(x_0)(x - x_0)\| \leq \varepsilon \cdot \|x - x_0\| = \|x - x_0\|.$$

By triangle inequality,

$$\|f(x) - f(x_0)\| \leq \|\mathbb{D}f(x_0)(x - x_0)\| + \|x - x_0\|.$$

Since  $\mathbb{D}f(x_0)$  is Lipschitz,  $\exists L$  s.t.

$$\|\mathbb{D}f(x_0)(x - x_0)\| \leq L \cdot \|x - x_0\|.$$

So,  $\|x - x_0\| < \delta \implies$

$$\begin{aligned} \|f(x) - f(x_0)\| &\leq L \cdot \|x - x_0\| + \|x - x_0\| \\ &= \underbrace{(L + 1)}_M \cdot \|x - x_0\| \\ &= M \cdot \|x - x_0\|. \end{aligned}$$

Q.E.D. ■

### Remark 6.6

- *Continuity is not sufficient to guarantee differentiability. For instance,  $f(x) = |x|$ .*

*However, differentiability  $\implies$  continuity.*

- *Derivative of a differentiable function may not be continuous. For example, consider the function*

$$f(x) = \begin{cases} x^2 \sin \frac{1}{x}, & x \neq 0 \\ 0, & x = 0. \end{cases}; f : \mathbb{R}^1 \rightarrow \mathbb{R}^1. \text{ Then, we have}$$

$$f'(0) = \lim_{x \rightarrow 0} \frac{f(x) - f(0)}{x} = \lim_{x \rightarrow 0} \frac{x^2 \sin \frac{1}{x}}{x} = \lim_{x \rightarrow 0} x \sin \frac{1}{x} = 0.$$

When  $x \neq 0$ ,

$$f'(x) = 2x \sin \frac{1}{x} + x^2 \cos \frac{1}{x} \left(-\frac{1}{x^2}\right) = 1x \sin \frac{1}{x} - \cos \frac{1}{x}.$$

*Conclusion:  $f$  is differentiable in  $\mathbb{R}^1$ . However,*

$$f'(x) = \begin{cases} 2x \sin \frac{1}{x} - \cos \frac{1}{x}, & x \neq 0 \\ 0, & x = 0 \end{cases}$$

*is not continuous at  $x = 0$ .*

**Theorem 6.2.3 Necessary Condition for Differentiability II**

Suppose  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  is differentiable. Then, the partial derivatives,  $\frac{\partial f_j}{\partial x_i}$ , exists  $\forall i, j$ .

**Example 6.2.4 The Converse is not True**

The converse of Theorem 6.2.3 is, in general, not true. Here we will consider a counterexample.

Consider function  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  given by

$$f(x, y) = \begin{cases} \frac{xy}{\sqrt{x^2 + y^2}}, & (x, y) \neq (0, 0) \\ 0, & (x, y) = (0, 0) \end{cases}$$

**Claim**  $f$  is continuous at  $(0, 0)$ .

In fact, we have  $(a - b)^2 \geq 0 \implies a^2 - 2ab + b^2 \geq 0$ . So,

$$ab \leq \frac{a^2 + b^2}{2} \quad a, b \in \mathbb{R}.$$

Then,

$$|xy| \leq \frac{1}{2}(a^2 + b^2) \implies \frac{xy}{\sqrt{x^2 + y^2}} \rightarrow 0 \quad \text{as } (x, y) \rightarrow (0, 0).$$

**Claim**  $\frac{\partial f(0, 0)}{\partial x} = 0$  and  $\frac{\partial f(0, 0)}{\partial y} = 0$ .

$$\frac{\partial f(0, 0)}{\partial x} = \lim_{x \rightarrow 0} \frac{f(x, 0) - f(0, 0)}{x - 0} = \lim_{x \rightarrow 0} \frac{0 - 0}{x} = 0.$$

**Claim**  $f$  is not differentiable at  $(0, 0)$ .

If  $f$  were differentiable, the matrix of  $\mathbb{D}f(0, 0)$  is given by

$$J_f(0, 0) = \left( \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right) = (0, 0).$$

However, note that

$$\frac{\|f(x, y) - f(0, 0) - \mathbb{D}f(x, y)\|}{\|(x, y) - (0, 0)\|} = \frac{\frac{|xy|}{\sqrt{x^2 + y^2}}}{\sqrt{x^2 + y^2}} = \frac{|xy|}{x^2 + y^2}.$$

Since  $\frac{|xy|}{x^2 + y^2}$  does not  $\rightarrow 0$  as  $(x, y) \rightarrow (0, 0)$ ,  $f$  is not differentiable at  $(0, 0)$ .

**Conclusion:** Continuity + Existence of Partial Derivative  $\frac{\partial f_j}{\partial x_i} \not\Rightarrow$  Differentiability.

### Theorem 6.2.5 Sufficient Condition for Differentiability

Let  $A \subset \mathbb{R}^n$  be open and  $f = (f_1, \dots, f_m) : A \rightarrow \mathbb{R}^m$ . If all the partials  $\frac{\partial f_j}{\partial x_i}$  exist and continuous on  $A$ , then  $f$  is differentiable on  $A$ .

**Proof2.** WTS:  $\forall x \in A$ ,

$$\lim_{y \rightarrow x} \frac{\|f(y) - f(x) - J_f(x)(y - x)\|}{\|y - x\|} = 0.$$

It is sufficient to show that this is true for each component  $f_j$  of  $f = (f_1, f_2, \dots, f_m)$ . Thus, we may assume  $m = 1$ :  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}^1$ . Then,

$$J_f(x) = \begin{pmatrix} \frac{\partial f}{\partial x_1} & \frac{\partial f}{\partial x_2} & \cdots & \frac{\partial f}{\partial x_n} \end{pmatrix}.$$

So,

$$J_f(y - x) = \sum_{i=1}^n \frac{\partial f}{\partial x_i} (y_i - x_i),$$

and

$$\begin{aligned} f(y) - f(x) &= f(y_1, y_2, \dots, y_n) - f(x_1, x_2, \dots, x_n) \\ &= f(y_1, y_2, \dots, y_n) - f(x_1, y_2, \dots, y_n) \\ &\quad + f(x_1, y_2, \dots, y_n) - f(x_1, x_2, \dots, y_n) \\ &\quad + f(x_1, x_2, \dots, y_n) - \cdots \quad \text{each time, we change one component} \\ &\quad + f(x_1, x_2, \dots, y_n) - f(x_1, x_2, \dots, x_n) \end{aligned}$$

By MVT,

$$f(y_1, y_2, \dots, y_m) - f(x_1, y_2, \dots, y_n) = \frac{\partial f}{\partial x_1} \cdot (y_1 - x_1).$$

Applying MVT to other terms, we obtain

$$f(y) - f(x) = \frac{\partial f(z^{(1)})}{\partial x_1} + \frac{\partial f(z^{(2)})}{\partial x_2} + \cdots + \frac{\partial f(z^{(n)})}{\partial x_n}.$$

Thus,

$$\begin{aligned} \|f(y) - f(x) - J_f(x)(y - x)\| &= \left\| \sum_{i=1}^n \frac{\partial f(z^{(i)})}{\partial x_i} - \sum_{i=1}^n \frac{\partial f(x)}{\partial x_i} \right\| \\ &\leq \sum_{i=1}^n \left| \frac{\partial f(z^{(i)})}{\partial x_i} - \frac{\partial f(x)}{\partial x_i} \right| \cdot \|y - x\| \end{aligned} \quad \begin{array}{l} \text{Triangle Inequality:} \\ |y_i - x_i| \leq \|y - x\| \end{array}$$

By continuity of partial derivative,  $\forall \varepsilon > 0, \exists \delta > 0$  s.t.

$$\|y - x\| < \delta \implies \sum_{i=1}^n \left| \frac{\partial f(z^{(i)})}{\partial x_i} - \frac{\partial f(x)}{\partial x_i} \right| < \varepsilon$$

Hence,

$$\|f(y) - f(x) - J_f(x)(y - x)\| < \varepsilon \|y - x\|.$$

Q.E.D. ■

**Definition 6.2.6 (Directional Derivative).** Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  and  $e \in \mathbb{R}^n$  be a unit vector. The directional derivative of  $f$  at  $x_0$  in the direction  $e$  is given by

$$D_e f(x_0) = \left. \frac{d}{dt} f(x_0 + te) \right|_{t=0} = \lim_{t \rightarrow 0} \frac{f(x_0 + te) - f(x_0)}{t}.$$

**Claim 6.2.7** If  $f$  is differentiable at  $x_0$ , then  $D_e f(x_0) = \mathbb{D}f(x_0) \cdot e$

**Proof 3.**

$$\begin{aligned} \lim_{t \rightarrow 0} \frac{\|f(x_0 + te) - f(x_0) - \mathbb{D}f(x_0)(te)\|}{\|te\|} &= 0 \\ \lim_{t \rightarrow 0} \frac{f(x_0 + te) - f(x_0)}{t} &= \mathbb{D}f(x_0)(e) \\ D_e f(x_0) &= \mathbb{D}f(x_0)(e). \end{aligned}$$

Q.E.D. ■

**Remark 6.7** Existence of directional derivatives  $\not\Rightarrow$  differentiability

#### Example 6.2.8

Continuity of  $f$  + Existence of directional derivative  $\not\Rightarrow$  differentiability.

Consider function  $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  given by

$$f(x, y) = \begin{cases} \frac{xy}{\sqrt{x^2 + y^2}}, & (x, y) \neq (0, 0) \\ 0, & (x, y) = (0, 0). \end{cases}$$

**Claim**  $D_e f(0, 0)$  exists for any direction  $e \in \mathbb{R}^2$ .

$$\lim_{t \rightarrow 0} \frac{f((0, 0) + te) - f(0, 0)}{t} \text{ exists } \forall e \in \mathbb{R}^2.$$

**Definition 6.2.9 (Tangent Line/Plane).**

- The *tangent line* to the curve  $y = f(x)$  at  $x_0$  is given by

$$y = f(x_0) + f'(x_0)(x - x_0).$$

- The *tangent plane* to the surface  $z = f(x)$  at  $x_0$  is given by

$$z = f(x_0) + \mathbb{D}f(x_0)(x - x_0).$$

**Example 6.2.10**

Find the tangent plane at  $(1, 2)$  to the surface  $z = x^2 + y^2$ .

**Solution 4.**

$$J_f(x) = \left( \frac{\partial f}{\partial x} \quad \frac{\partial f}{\partial y} \right) = (2x \quad 2y).$$

The tangent plane is given by

$$\begin{aligned} z &= f(1, 2) + \mathbb{D}f(1, 2)((x, y) - (1, 2)) \\ &= 1^2 + 2^2 + \left[ 2x \quad 2y \right] \Big|_{(x,y)=(1,2)} \begin{bmatrix} x - 1 \\ y - 2 \end{bmatrix} \\ &= 5 + \begin{bmatrix} 2 & 4 \end{bmatrix} \begin{bmatrix} x - 1 \\ y - 2 \end{bmatrix} \\ z &= 5 + 2(x - 1) + 4(y - 2). \end{aligned}$$

□

**Summary III: Relations among Properties of  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$** 

Differentiability

 $\implies$  Existence of Directional Derivative $\implies$  Existence of Partial Derivative (moving in direction of the basis)+ **Theorem 6.2.5**  $\implies$  Differentiability $\nRightarrow$  Continuity $\implies$  Continuity $\nRightarrow$  Existence of Partial Derivative**6.3 Differentiation Rules****6.3.1 Chain Rule**Recall the one variable case:  $h = g(u)$ ,  $u = f(x)$ . Then,

$$h = f \circ f(x) = g(f(x)),$$

and

$$\frac{dh}{dx} = \frac{dh}{du} \cdot \frac{du}{dx} = g'(f(x)) \cdot f'(x).$$

**Theorem 6.3.1 General Case Chain Rule**

Let  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  and  $g : B \rightarrow \mathbb{R}^p$  be differentiable with  $f(A) \subset B$ . Then, the composite  $g \circ f : A \rightarrow \mathbb{R}^p$  is differentiable, and

$$\mathbb{D}(g \circ f)(x) = \mathbb{D}g(f(x)) \circ \mathbb{D}f(x),$$

a composition of linear mappings.

In matrix notation, define  $h = g(u)$  and  $u = f(x)$ . Then,  $h = g \circ f(x) = g(f(x))$ , and

$$J_h(x) = J_g(f(x)) \cdot J_f(x) \quad \text{product of matrices}$$

$$= \begin{bmatrix} \frac{\partial g_1}{\partial u_1} & \cdots & \frac{\partial g_1}{\partial u_m} \\ \vdots & \ddots & \vdots \\ \frac{\partial g_p}{\partial u_1} & \cdots & \frac{\partial g_p}{\partial u_m} \end{bmatrix} \cdot \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

**Proof 1.** (Sketch). We need to show: for fixed  $x \in A \subset \mathbb{R}^n$ ,

$$\lim_{y \rightarrow x} \frac{\|h(y) - h(x) - \mathbb{D}h(x)(y - x)\|}{\|y - x\|} = 0,$$



or

$$\lim_{y \rightarrow x} \frac{\|g(f(y)) - g(f(x)) - \mathbb{D}g(f(x))[\mathbb{D}f(x)(y - x)]\|}{\|y - x\|} = 0.$$

Work with the numerator:

$$\begin{aligned} \text{numerator} &= \|g(f(y)) - g(f(x)) - \mathbb{D}g(f(x))(f(y) - f(x)) \\ &\quad + \mathbb{D}g(f(x))(f(y) - f(x)) - \mathbb{D}g(f(x))[\mathbb{D}f(x)(y - x)]\| \\ &\leq \|g(f(y)) - g(f(x)) - \mathbb{D}g(f(x))[\mathbb{D}f(x)(y - x)]\| && \text{triangle inequality} \\ &\quad + \|\mathbb{D}g(f(x))(f(y) - f(x)) - \mathbb{D}g(f(x))[\mathbb{D}f(x)(y - x)]\| \\ &\leq \varepsilon_1 \|f(y) - f(x)\| + \|\mathbb{D}g(f(x))\| \cdot \|f(y) - f(x) - \mathbb{D}f(x)(y - x)\| \\ &\quad (\varepsilon_1 : g \text{ is differentiable; } \mathbb{D}g(f(x)) : \text{common factor}) \\ &\leq \varepsilon_1 \cdot L \|y - x\| + M \cdot \varepsilon_2 \|y - x\| \\ &\quad (L : \text{local Lipschitz; } M : \text{differential bounded; } \varepsilon_2 : f \text{ is differentiable}) \\ &= (L\varepsilon_1 + M\varepsilon_2) \cdot \|y - x\|. \end{aligned}$$

Therefore,

$$\lim_{y \rightarrow x} \frac{\text{numerator}}{\|y - x\|} = \lim_{y \rightarrow x} \frac{(L\varepsilon_1 + M\varepsilon_2)\|y - x\|}{\|y - x\|} = \lim_{y \rightarrow x} L\varepsilon_1 + M\varepsilon_2 = 0.$$

Q.E.D. ■

### Example 6.3.2

- Change of Variable

$$(x, y, z) \longleftrightarrow (r, \theta, z) : \begin{cases} x = r \cos \theta \\ y = r \sin \theta \\ z = z \end{cases} \quad (\text{cylindrical coordinate})$$

Let  $h(r, \theta, z) = f(x, y, z) = f(x(r, \theta, z), y(r, \theta, z), z(r, \theta, z))$ . Then,

$$\mathbb{D}h = \frac{\partial h}{\partial(r, \theta, z)} = \frac{\partial f}{\partial(x, y, z)} \cdot \frac{\partial(x, y, z)}{\partial(r, \theta, z)} = J_f \cdot \begin{bmatrix} \cos \theta & -r \sin \theta & 0 \\ \sin \theta & r \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- Consider composition of the maps  $[0, 1] \xrightarrow{\gamma} \mathbb{R}^n \xrightarrow{f} \mathbb{R}$ . Then,  $h(t) = f(\gamma(t))$ . By chain rule,

$$\begin{aligned} h'(t) &= \mathbb{D}f \circ \mathbb{D}\gamma = \begin{pmatrix} \frac{\partial f}{\partial x_1} & \frac{\partial f}{\partial x_2} & \cdots & \frac{\partial f}{\partial x_n} \end{pmatrix} \begin{pmatrix} x'_1(t) \\ x'_2(t) \\ \vdots \\ x'_n(t) \end{pmatrix} \\ &= \sum_{i=1}^n \frac{\partial f}{\partial x_i} x'_i(t) = \langle \nabla f, \gamma'(t) \rangle. \end{aligned}$$

### 6.3.2 Other Differentiation Rules

#### Theorem 6.3.3 Product Rule

Let  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}^m$  and  $g : A \rightarrow \mathbb{R}$  be differentiable. Then, the product  $gf : A \rightarrow \mathbb{R}^m$  is differentiable, and

$$\mathbb{D}(gf) = g(\mathbb{D}f) + (\mathbb{D}g)f.$$

More precisely, for each  $x \in A$  and  $h \in \mathbb{R}^n$ ,

$$\mathbb{D}(gf) \cdot h = \underbrace{g(x)}_{\text{scalar}} \cdot \underbrace{\mathbb{D}f(x)(h)}_{\in \mathbb{R}^m} + \underbrace{\mathbb{D}g(x)(h)}_{\text{scalar}} \cdot \underbrace{f(x)}_{\in \mathbb{R}^m}.$$

In particular,

$$\frac{\partial g(f_j)}{\partial x_i} = g \cdot \frac{\partial f_j}{\partial x_i} + \frac{\partial g}{\partial x_i} \cdot f_j.$$

#### Theorem 6.3.4 Other Differentiation Rules

$$\mathbb{D}(f + g) = \mathbb{D}f + \mathbb{D}g$$

$$\mathbb{D}(\lambda f) = \lambda \mathbb{D}f$$

$$\mathbb{D}\left(\frac{f}{g}\right) = \frac{g\mathbb{D}f - (\mathbb{D}g)f}{g^2} \quad \left(\text{derived from product rule: } \frac{f}{g} = f \cdot \frac{1}{g}\right)$$

## 6.4 Geometric Interpretation of Gradient

Let  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable.

**Definition 6.4.1**  $(\mathbb{D}f(x), \nabla f(x), D_{\text{eff}}(x))$ .

- Differential of  $f$ : a matrix/linear transformation

$$\mathbb{D}f(x) = \begin{bmatrix} \frac{\partial f}{\partial x_1} & \frac{\partial f}{\partial x_2} & \cdots & \frac{\partial f}{\partial x_n} \end{bmatrix}$$

- Gradient of  $f$ : a vector

$$\nabla f(x) = \left( \frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}, \dots, \frac{\partial f}{\partial x_n} \right).$$

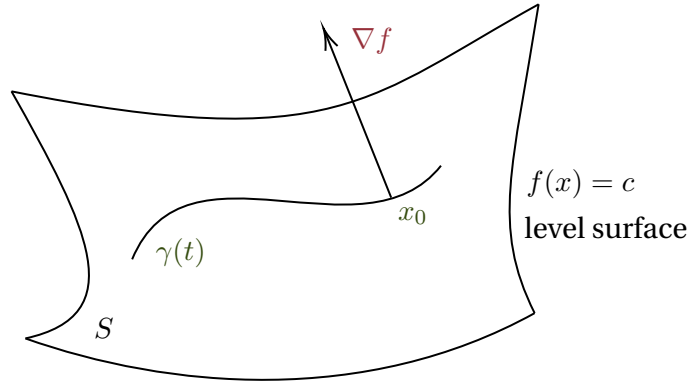
- Directional derivative of  $f$  in the direction  $e$ :

$$D_e f(x) = \mathbb{D}f(x)e = \langle \nabla f(x), e \rangle.$$

Geometric meaning of  $D_e f(x)$ : Rate of change in the direction of  $e$ .

### 6.4.2 Geometric Meaning of Gradient.

**Claim 6.4.3**  $\nabla f$  is perpendicular to the level surface  $S$  defined by  $f(x) = \text{constant}$ .



**Proof 1.** Fix any curve  $\gamma(t)$  on  $S$ :  $\gamma : [a, b] \rightarrow S$ . Then,  $f(\gamma(t)) = c$ . By chain rule,

$$\mathbb{D}f(\gamma(t)) \cdot \gamma'(t) = 0 \implies \langle \nabla f(x_0), \gamma'(x_0) \rangle = 0.$$

So,  $\nabla f(x_0) \perp \gamma'(x_0)$ . That is,  $\nabla f \perp \text{curve } \gamma \text{ on } S \implies \nabla f \perp S$ .

Q.E.D. ■

**Corollary 6.4.4 Tangent Plane:** The tangent plane at  $x_0$  of the level surface is given by

$$\langle \nabla f(x_0), x - x_0 \rangle = 0.$$

#### Example 6.4.5

Find the tangent plane at  $(1, 0, 1)$  to the surface  $x^2 - y^2 + xyz = 1$ .

**Claim 6.4.6** The direction of  $\nabla f$  is the direction in which  $f$  has the greatest rate of change, which is given by  $\|\nabla f\|$ .

**Proof2.** Fix a direction  $e \in \mathbb{R}^n$ . Then, the rate of change in direction  $e$  is given by

$$D_e f(x_0) = \langle \nabla f, e \rangle = \|\nabla f\| \|e\| \cos \theta,$$

where  $\theta$  is the angle between  $\nabla f(x_0)$  and  $e$ . Then, the rate of change is maximized when  $\cos \theta = 1$ . So,  $\theta = 0$ . That is,  $e$  is in the direction of  $\nabla f$ .

Q.E.D. ■

## 6.5 Mean Value Theorem (MVT)

### Theorem 6.5.1 MVT in 1-D

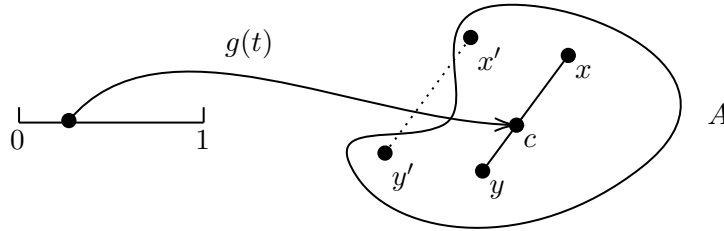
Let  $f : [a, b] \rightarrow \mathbb{R}^1$  be continuous and differentiable on  $(a, b)$ . Then,  $\exists c \in (a, b)$  s.t.

$$f'(c) = \frac{f(b) - f(a)}{b - a} \quad \text{or} \quad f(b) - f(a) = f'(c)(b - a).$$

### Theorem 6.5.2 MVT in Higher Dimension

Let  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be differentiable on an open set  $A$ . Then, for any pair of points  $x, y \in A$  s.t. the line segment  $[x, y]$  joining  $x$  and  $y$  is contained in  $A$ ,  $\exists$  a point  $c \in [x, y]$  s.t.

$$f(y) - f(x) = \mathbb{D}f(c)(y - x).$$



**Proof1.** Let  $g(t) = (1 - t)x + ty$  for  $0 \leq t \leq 1$  and

$$h(t) = f \circ g(t) = f((1 - t)x + ty) : [0, 1] \rightarrow \mathbb{R}.$$

Apply Theorem 6.5.1 to  $h$ , we know  $\exists t_0 \in (0, 1)$  s.t.

$$\begin{aligned} h(1) - h(0) &= h'(t_0)(1 - 0) \\ f(y) - f(x) &= \mathbb{D}f(g(t_0)) \cdot g'(t_0) && \text{[Chain Rule]} \\ &= \mathbb{D}f(g(t_0)) \cdot (y - x). \end{aligned}$$

Denote  $g(t_0) = c \in [x, y]$ . Then,

$$f(y) - f(x) = \mathbb{D}f(c)(y - x).$$

Q.E.D. ■

**Definition 6.5.3 (Convex Set).** A set  $A \subset \mathbb{R}^n$  is *convex* if  $\forall x, y \in A, [x, y] \subset A$ .

**Corollary 6.5.4 :** Let  $A \subset \mathbb{R}^n$  be open and convex, and  $f : A \rightarrow \mathbb{R}^m$  differentiable. If  $\mathbb{D}f \equiv 0$ , then  $f$  is constant in  $A$ .

**Proof 2.** (Sketch)

Apply MVT to each component of  $f = (f_1, f_2, \dots, f_m)$ .

Q.E.D. ■

## 6.6 Taylor's Theorem & Higher Order Differentials

### 6.6.1 One Dimensional Case

#### Theorem 6.6.1 Taylor's Formula

Let  $f : (a, b) \rightarrow \mathbb{R}$  be one of class  $C^r$  (i.e.,  $f'(x), f''(x), \dots, f^{(r)}(x)$  are continuous). Then, for any  $x_0, x \in (a, b)$ , we have

$$f(x) = \underbrace{f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2!}(x - x_0)^2 + \dots + \frac{f^{(r-1)}(x_0)}{(r-1)!}(x - x_0)^{r-1}}_{\text{Taylor's polynomial of degree } r-1} + \overbrace{R_{r-1}(x_0)}^{\text{Remainder}},$$

where  $R_{r-1}$  is the remainder at  $x_0$  and can be written as

$$R_{r-1}(x_0) = \frac{f^{(r)}(c)}{r!}(x - x_0)^r \quad \text{for some } c \text{ between } x \text{ and } x_0.$$

**Remark 6.8 (Key Idea to Prove)** Use integration by parts in a reversed way multiple times.

**Proof 1.** Write  $h = x - x_0$ . Then, by Fundamental Theorem of Calculus,

$$f(x) - f(x_0) = f(x_0 + h) - f(x_0) = \int_0^1 \frac{d}{dt} f(x_0 + th) dt.$$

Now, apply integration by parts. Taking  $u = \frac{d}{dt} f(x_0 + th) = f'(x_0 + th)h$  and  $dv = dt \implies v = t - 1$ , we have

$$\begin{aligned} f(x) - f(x_0) &= \int_0^1 u dv \\ &= uv \Big|_0^1 - \int_0^1 v du \\ &= -(-1)f'(x_0)h - \int_0^1 (t-1)f''(x_0 + th)h^2 dt \\ &= f'(x_0)h - \int_0^1 f''(x_0 + th)h^2(t-1) dt. \end{aligned}$$

Apply integration by parts again with

$$u = f''(x_0 + th)h^2 \quad \text{and} \quad dv = (t - 1) dt \implies v = \frac{1}{2}(t - 1)^2.$$

Then, we obtain

$$\begin{aligned} \int_0^1 f''(x_0 + th)h^2(t - 1) dt &= f''(x_0 + th)h^2 \frac{1}{2}(t - 1)^2 \Big|_0^1 - \int_0^1 \frac{1}{2}(t - 1)^2 f'''(x_0 + th)h^3 dt \\ &= \frac{f''(x_0)}{2}h^2 + \int_0^1 f^{(3)}(x_0 + th)h^3 \cdot \frac{1}{2}(t - 1)^2 dt. \end{aligned}$$

So,

$$f(x) - f(x_0) = f'(x_0)h + \frac{f''(x_0)}{2}h^2 + \int_0^1 f^{(3)}(x_0 + th)h^3 \cdot \frac{1}{2}(t - 1)^2 dt.$$

By induction, we obtain that

$$\begin{aligned} f(x) - f(x_0) &= \underbrace{f'(x_0)h + \frac{f''(x_0)}{2}h^2 + \frac{f'''(x_0)}{3!}h^3 + \cdots + \frac{f^{(r-1)}(x_0)}{(r-1)!}h^{r-1}}_{\text{Taylor's polynomial}} \\ &\quad + \underbrace{(-1)^{r-1} \int_0^1 f^{(r)}(x_0 + th)h^r \frac{(t-1)^{r-1}}{(r-1)!} dt}_{\text{Remainder}} \end{aligned}$$

**Lemma 6.6.2 2<sup>nd</sup> MVT for Integral:** If  $g \geq 0$ , then  $\int_a^b f(x)g(x) dx = f(\lambda) \int_a^b g(x) dx$ .

Apply 2<sup>nd</sup> MVT to the remainder, we have

$$\begin{aligned} R_{r-1} &= (-1)^{r-1} f^{(r)}(x_0 + t_0 h)h^r \int_0^1 \frac{(t-1)^{r-1}}{(r-1)!} dt \\ &= f^{(r)}(x_0 + t_0 h)h^r \cdot \frac{1}{r} \quad \left[ (-1)^{r-1} \text{ is absorbed when evaluating the integral} \right] \\ &= \frac{f^{(r)}(c)}{r!}h^r \quad \left[ \text{Denote } c = x_0 + t_0 h, \text{ a point between } x_0 \text{ and } x \right] \end{aligned}$$

Combining everything, we get exactly what we have claimed.

Q.E.D. ■

#### Summary IV: Taylor's Formula & Taylor's Approximation

- Taylor's Formula:

$$f(x) = P_{r-1}(x) + R_{r-1}.$$

- Taylor's Approximation:

$$f(x) \approx P_{r-1}(x).$$

### 6.6.2 Taylor Series

**Definition 6.6.3 (Taylor Series).** Let  $f \in C^\infty$ . Then, the *Taylor series* is defined as

$$\sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} (x - x_0)^n = f(x_0) + f'(x_0)(x - x_0) + \frac{f''(x_0)}{2}(x - x_0)^2 + \cdots .$$

**Definition 6.6.4 (Real Analytic).**  $f$  is (real) *analytic* at  $x_0$  if its Taylor series converges to  $f(x)$  in a neighborhood of  $x_0$ . i.e.,

$$f(x) = \sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} (x - x_0)^n, \quad |x - x_0| < R.$$

**Corollary 6.6.5 :** If  $f \in C^\infty(\mathbb{R})$  and for each interval  $[a, b]$ ,  $\exists$  constant  $M$  s.t.

$$|f^{(n)}(x)| \leq M^n \quad \forall n \text{ and } x \in [a, b],$$

then,  $f$  is real analytic at each point  $x_0$  and it has Taylor series representation. Namely,

$$f(x) = \sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} (x - x_0)^n, \quad |x - x_0| < \infty.$$

**Proof 2.** Fix  $x_0 \in \mathbb{R}$ . For any  $x \in \mathbb{R}$ , choose  $b > 0$  s.t.  $x_0, x \in [-b, b]$ . By Taylor's Formula,

$$f(x) = \underbrace{P_{n-1}(x)}_{\substack{\text{partial sum} \\ \text{of the series}}} + R_{n-1}.$$

Recall:

$$R_{n-1} = \frac{f^{(n)}(c)}{n!} (x - x_0)^n \quad \text{for some } c.$$

Then,

$$|R_{n-1}| \leq \frac{M^n}{n!} |x - x_0|^n \quad \forall x \in [-b, b].$$

Since the series  $\sum_{n=0}^{\infty} \frac{M^n}{n!} |x - x_0|^n$  converges by ratio test, its  $n$ -th term,

$$\frac{M^n}{n!} |x - x_0|^n \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Hence,  $R_{n-1} \rightarrow 0$  as  $n \rightarrow \infty$ . Then,  $P_{n-1}(x) \rightarrow f(x)$  as  $n \rightarrow \infty$ .

Q.E.D. ■

#### Example 6.6.6

- $e^x$  and  $\sin x$  are real analytic in  $\mathbb{R}$ . Find Taylor series at  $x_0 = 0$ :

**Solution 3.**

$$e^n = \sum_{n=0}^{\infty} \frac{1}{n!} x^n, \quad |x - x_0| < \infty.$$

□

- Is every  $C^\infty$  real analytic? No.

**Counterexample 6.7.** Consider the function  $f(x)$ :

$$f(x) = \begin{cases} 0, & x = 0 \\ e^{-1/x^2}, & x \neq 0. \end{cases}$$

**Claim**  $f(x) \in C^\infty$ .

*Proof.* At  $x = 0$ ,

$$f'(0) = \lim_{x \rightarrow 0} \frac{f(x) - f(0)}{x - 0} = \lim_{x \rightarrow 0} \frac{e^{-1/x^2}}{x} = 0 \quad (\text{by L.H.})$$

At  $x \neq 0$ ,

$$f'(x) = e^{-1/x^2} \left( \frac{2}{x^3} \right) = \frac{2/x^3}{e^{1/x^2}} \rightarrow 0 \quad \text{as } x \rightarrow 0 \quad (\text{by L.H.})$$

So,  $f'(x)$  is continuous at  $x = 0$ , and

$$f'(x) = \begin{cases} 0, & x = 0 \\ e^{-1/x^2} \left( \frac{2}{x^3} \right) & x \neq 0. \end{cases}$$

By induction, one can show that

- $f^{(n)}(0) = 0 \quad \forall n$
- $f^{(n)}(x) \rightarrow 0$  as  $x \rightarrow 0$ .

So,  $f^{(n)}(x)$  is continuous at  $x = 0$ . So,  $f \in C^\infty$ . □

**Claim**  $f(x)$  is not real analytic at  $x = 0$ .

*Proof.* Taylor series:

$$\sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} (x - x_0)^n = \sum_{n=0}^{\infty} \frac{f^{(n)}(0)}{n!} (x)^n = 0.$$

So, the Taylor series does not converge to  $f(x)$  on any neighborhood of  $x = 0$ . □



### 6.6.3 Higher Dimensional Case

**Observation:** Let  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$ .

- **Differential:**  $\mathbb{D}f(x)$  is a linear transformation  $\mathbb{R}^n \rightarrow \mathbb{R}$ .
- Let  $g(x) = \mathbb{D}f(x)$ . Then,  $g : A \subset \mathbb{R}^n \rightarrow \mathbf{L}(\mathbb{R}^n, \mathbb{R}) \approx \mathbb{R}^n$ , where  $\mathbf{L}(M, N)$  is the space of linear transformation from  $M$  to  $N$ .
- $\mathbb{D}g(x)$  is a linear transformation  $\mathbb{R}^n \rightarrow \mathbb{R}^n$  or  $\mathbf{L}(\mathbb{R}^n, \mathbb{R})$ .

**Notation 6.8.** Higher Order Differential The second order differential of  $f$  at  $x$  is denoted as

$$\mathbb{D}^2 f(x) = \mathbb{D}g(x) = \mathbb{D}(\mathbb{D}f(x)).$$

**Definition 6.6.9 (Bilinear Maps).** Given  $f$  and  $x \in A$ . Define a *bilinear map*,  $\mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$  by

$$\mathbb{D}^2 f(x)(u, v) = \left[ \mathbb{D}^2 f(x)(u) \right] (v),$$

where  $u, v \in \mathbb{R}^n$  and  $\mathbb{D}^2 f(x)(u) \in \mathbf{L}(\mathbb{R}^n, \mathbb{R})$ . In matrix notation,

$$uBv^\top,$$

where  $u$  is  $1 \times n$ ,  $B$  is  $n \times n$ , and  $v^\top$  is  $n \times 1$ .

**Definition 6.6.10 (Matrix Representation of the Bilinear Map).**  $\mathbb{D}^2 f(x) : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$  is given by

$$\begin{bmatrix} \frac{\partial^2 f}{\partial x_1 \partial x_1} & \frac{\partial^2 f}{\partial x_2 \partial x_1} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_1} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_1 \partial x_n} & \frac{\partial^2 f}{\partial x_2 \partial x_n} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_n} \end{bmatrix}_{n \times n}$$

This matrix is denoted as  $H_x(f)$ , the Hessian matrix of  $f$  at  $x$ . Then, in matrix form, we have that for  $u = (u_1, u_2, \dots, u_n) \in \mathbb{R}^n$  and  $v = (v_1, v_2, \dots, v_n) \in \mathbb{R}^n$ , and

$$\mathbb{D}^2 f(x)(u, v) = u \cdot H_x(f) \cdot v^\top \in \mathbb{R}.$$

**Proof 4.** Note that

$$g(x) = \left( \frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2}, \dots, \frac{\partial f}{\partial x_n} \right) : \mathbb{R}^n \rightarrow \mathbb{R}^n.$$

Then,

$$\begin{aligned}\mathbb{D}^2 f(x) &= \mathbb{D}g(x) \\ &= \begin{bmatrix} \frac{\partial}{\partial x_1} \frac{\partial f}{\partial x_1} & \frac{\partial}{\partial x_2} \frac{\partial f}{\partial x_1} & \cdots & \frac{\partial}{\partial x_n} \frac{\partial f}{\partial x_1} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial}{\partial x_1} \frac{\partial f}{\partial x_n} & \frac{\partial}{\partial x_2} \frac{\partial f}{\partial x_n} & \cdots & \frac{\partial}{\partial x_n} \frac{\partial f}{\partial x_n} \end{bmatrix}.\end{aligned}$$

Q.E.D. ■

**Lemma 6.6.11 Symmetry of the Partial and Differentials:** Let  $f(x, y) : A \subset \mathbb{R}^2 \rightarrow \mathbb{R}$  be of class  $\mathcal{C}^2$ . Then,

$$\frac{\partial^2 f}{\partial x \partial y} = \frac{\partial^2 f}{\partial y \partial x}.$$

In general, for  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  in class  $\mathcal{C}^2$ ,

$$\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i} \quad \forall i, j.$$

**Extension 6.1** If  $f \in \mathcal{C}^{(n)}$ , the order of taking  $n$ -th derivative does not matter.

**Corollary 6.6.12 :** If  $f$  is of class  $\mathcal{C}^2$ , then  $\mathbb{D}^2 f(x) : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$  is symmetric. That is,

$$\mathbb{D}^2 f(x)(u, v) = \mathbb{D}^2 f(x)(v, u).$$

**Proof 5.**

$$\mathbb{D}^2 f(x)(u, v) = u \cdot H_x(f) \cdot v^\top$$

Since  $\mathbb{D}^2 f(x)(u, v) \in \mathbb{R}$ , we have

$$\begin{aligned}\mathbb{D}^2 f(x)(u, v) &= [\mathbb{D}^2 f(x)(u, v)]^\top = (u \cdot H_x(f) \cdot v^\top)^\top \\ &= v \cdot H_x(f)^\top \cdot u^\top \\ &= v \cdot H_x(f) \cdot u^\top && \text{[by symmetry of } H_x(f)\text{]} \\ &= \mathbb{D}^2 f(x)(u, v).\end{aligned}$$

Q.E.D. ■

### Example 6.6.13 Symmetry of Partial

Let  $f(x, y, z) = e^{xyz} + xyz : \mathbb{R}^3 \rightarrow \mathbb{R}$ . Verify the symmetry of the partials.

**Solution 6.**

$$\frac{\partial f}{\partial x} = ye^{xy} + yz; \quad \frac{\partial f}{\partial y} = xe^{xy} + yz; \quad \frac{\partial f}{\partial z} = xy.$$

$$\frac{\partial^2 f}{\partial y \partial x} = \frac{\partial}{\partial y} \left( \frac{\partial f}{\partial x} \right) = e^{xy} + xy e^{xy} + z;$$

$$\frac{\partial^2 f}{\partial x \partial y} = \frac{\partial}{\partial x} \left( \frac{\partial f}{\partial y} \right) = e^{xy} + xy e^{xy} + z.$$

□

### Summary V: Higher Order Differentials

- 1-st Order Differential:  $\mathbb{D}f(x_0) : \mathbb{R}^n \rightarrow \mathbb{R}$ : 1-linear form

$$\mathbb{D}f(x_0)(v) = J_f(x_0) \cdot v = \sum_{i=1}^n \frac{\partial f}{\partial x_i} v_i.$$

- 2-nd Order Differential:  $\mathbb{D}^2 f(x_0) : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ : bilinear form

$$\mathbb{D}^2 f(x_0)(v, w) = v \cdot H_f(x_0) \cdot w^\top = \sum_{i,j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j} v_i \cdot w_j.$$

- $k$ -th Order Differential:  $\mathbb{D}^k f(x_0) : \mathbb{R}^n \times \mathbb{R}^n \times \cdots \times \mathbb{R}^n \rightarrow \mathbb{R}$ :  $k$ -linear form

$$\mathbb{D}^k f(x_0)(v^{(1)}, v^{(2)}, \dots, v^{(k)}) = \sum_{i_1, i_2, \dots, i_k=1}^n \frac{\partial^k f}{\partial x_{i_1} \partial x_{i_2} \cdots \partial x_{i_k}} v_{i_1}^{(1)} v_{i_2}^{(2)} \cdots v_{i_k}^{(k)}$$

In particular, denote  $h = x - x_0 \in \mathbb{R}^n$ , then

$$\mathbb{D}^k f(x_0)(h, h, \dots, h) = \sum \frac{\partial^k f}{\partial x_{i_1} \partial x_{i_2} \cdots \partial x_{i_k}} h_{i_1} h_{i_2} \cdots h_{i_k}.$$

- Speical case:  $n = 2$ : Write  $\mathbb{D}^k f(x_0)(h, h) = \left( \frac{\partial}{\partial x_1} + \frac{\partial}{\partial x_2} \right)^k f(x_0) \cdot (h, h)$ . Then,

$$\mathbb{D}^1 f = \left( \frac{\partial}{\partial x_1} + \frac{\partial}{\partial x_2} \right)^1 f = \frac{\partial f}{\partial x_1} + \frac{\partial f}{\partial x_2}; \quad \mathbb{D}^2 f = \left( \frac{\partial}{\partial x_1} + \frac{\partial}{\partial x_2} \right)^2 f = \frac{\partial^2 f}{\partial x_1^2} + 2 \frac{\partial^2 f}{\partial x_1 \partial x_2} + \frac{\partial^2 f}{\partial x_2^2},$$

$$\mathbb{D}^3 f(h, h, h) = \frac{\partial^3 f}{\partial x_1^3} h_1^3 + 3 \frac{\partial^3 f}{\partial x_1^2 \partial x_2} h_1^2 h_2 + 3 \frac{\partial^3 f}{\partial x_1 \partial x_2^2} h_1 h_2^2 + \frac{\partial^3 f}{\partial x_2^3} h_2^3$$

**Theorem 6.6.14 Taylor's Theorem**

Let  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be of class  $\mathcal{C}^r$ . Suppose  $x, x_0 \in A$  s.t. the line segment joining  $x$  and  $x_0$ ,  $[x, x_0] \subset A$ . Then,  $\exists c \in [x, x_0]$  s.t.

$$f(x) = f(x_0) + \mathbb{D}f(x_0)(x - x_0) + \frac{1}{2!}\mathbb{D}^2f(x_0)(x - x_0, x - x_0) + \cdots \\ + \frac{1}{(r-1)!}\mathbb{D}^{r-1}f(x_0)(x - x_0, x - x_0, \dots, x - x_0) + R_{r-1},$$

where  $R_{r-1}$  is the remainder given by

$$R_{r-1} = \frac{1}{r!}\mathbb{D}^r f(c)(x - x_0, \dots, x - x_0)$$

and satisfies

$$\frac{R_{r-1}(x_0)}{\|x - x_0\|^{r-1}} \rightarrow 0 \quad \text{as } x \rightarrow x_0.$$

**Proof 7.** Consider 1-variable form,  $\varphi(t) = x_0 + t(x - x_0)$ . Define

$$g(t) = f(x_0 + t(x - x_0))$$

for  $t \in (a, b)$  with  $[0, 1] \subset (a, b)$ .

Apply Taylor's Theorem in 1-D to  $g(t)$ , we get

$$g(1) = g(0) + g'(0)(1 - 0) + \frac{g''(0)}{2!}(1 - 0)^2 + \cdots + \frac{g^{(r-1)}(0)}{(r-1)!}(1 - 0)^{r-1} + R_{r-1} \\ f(x) = f(x_0) + \sum_{k=1}^{r-1} \frac{g^{(k)}(0)}{k!} + \frac{1}{r!}g^{(r)}(\tilde{c}), \quad \tilde{c} \in [0, 1].$$

By chain rule, one can get

$$g'(t) = \mathbb{D}f(\varphi(t))\varphi'(t) \\ g'(0) = \mathbb{D}f(x_0)(x - x_0) \\ g''(t) = \mathbb{D}^2f(\varphi(t))\varphi'(t) \cdot \varphi'(t) \\ g''(0) = \mathbb{D}^2f(x_0)(x - x_0)^2 = \mathbb{D}^2f(x_0)(x - x_0, x - x_0).$$

So,

$$g^{(k)}(0) = \mathbb{D}^k f(x_0)(x - x_0, x - x_0, \dots, x - x_0).$$

Q.E.D. ■

**Example 6.6.15 Polynomial Approximation using Taylor's Theorem**

Determine the 2-nd order Taylor's formula for  $f(x, y) = e^{(x-1)^2} \cos y$  at  $(1, 0)$ .

**Solution 8.**

- Compute partials:

$$\frac{\partial f}{\partial x} = 2(x-1)e^{(x-1)^2} \cos y; \quad \frac{\partial f}{\partial y} = -e^{(x-1)^2} \sin y.$$

$$\frac{\partial^2 f}{\partial x^2} = 2e^{(x-1)^2} \cos y + 4(x-1)^2 e^{(x-1)^2} \cos y; \quad \frac{\partial^2 f}{\partial y^2} = -e^{(x-1)^2} \cos y.$$

$$\frac{\partial^2 f}{\partial x \partial y} = -2(x-1)e^{(x-1)^2} \sin y$$

- Evaluate at base point  $(1, 0)$ :

$$\left. \frac{\partial f}{\partial x} \right|_{(1,0)} = 0, \quad \left. \frac{\partial f}{\partial y} \right|_{(1,0)} = 0, \quad \left. \frac{\partial^2 f}{\partial x^2} \right|_{(1,0)} = 2, \quad \left. \frac{\partial^2 f}{\partial x \partial y} \right|_{(1,0)} = 0, \quad \left. \frac{\partial^2 f}{\partial y^2} \right|_{(1,0)} = 1.$$

- Taylor's Formula:  $h = x - x_0 = (x, y) - (1, 0)$ .

$$f(x, y) = f(1, 0) + \mathbb{D}f(1, 0)(h) + \mathbb{D}^2 f(1, 0)(h, h) + R_2,$$

where  $f(1, 0) = 1$ ,  $\mathbb{D}f(1, 0) = \begin{bmatrix} 0 & 0 \end{bmatrix}$ , and  $\mathbb{D}^2 f(1, 0) = \begin{bmatrix} 2 & 0 \\ 0 & -1 \end{bmatrix}$ . So,

$$\mathbb{D}f(1, 0)(h) = 0$$

$$\mathbb{D}^2 f(1, 0)(h, h) = (x-1, y) \begin{bmatrix} 2 & 0 \\ 0 & -1 \end{bmatrix} \begin{pmatrix} x-1 \\ y \end{pmatrix} = 2(x-1)^2 - y^2.$$

Then,

$$f(x, y) = 1 + \frac{1}{2} [2(x-1)^2 - y^2] + R_2,$$

where

$$\frac{R_2}{\|(x-1, y)\|^2} \rightarrow 0 \quad \text{as} \quad (x-1, y) \rightarrow (1, 0).$$

□

**6.7 Minima & Maxima in  $\mathbb{R}^n$** 

**Question:** Given function  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$ , how do we find (local) maximum or minimum points for  $f$  in  $A$ ?

**6.7.1 Optimization in 1-D.** Suppose  $f : (a, b) \rightarrow \mathbb{R}$ 

- A local max/min point (or extreme point)  $x_0$  must be a critical point:

$$f'(x_0) = 0 \quad \text{or} \quad f'(x_0) \text{ D.N.E.}$$

- 2-nd Order Derivative Test (for critical points):

$$f''(x_0) > 0 : \text{local min}; \quad f''(x_0) < 0 : \text{local max.}$$

**Definition 6.7.2 (Extrema).** Suppose  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$ .

- Then,  $x_0 \in A$  is a *local minimum* if  $\exists \delta > 0$  s.t.  $x \in A$  and

$$|x - x_0| < \delta \implies f(x) \geq f(x_0).$$

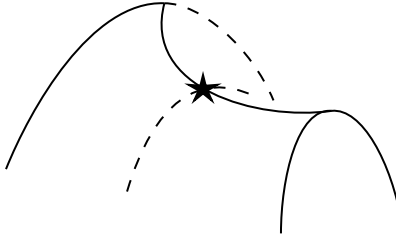
- Similarly,  $x_0 \in A$  is a *local maximum* if  $\exists \delta > 0$  s.t.  $x \in A$  and

$$|x - x_0| < \delta \implies f(x) \leq f(x_0).$$

**Theorem 6.7.3 Necessary Condition for Extreme Points**

If  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  is differentiable and  $x_0 \in A$  is an extreme point for  $f$ , then  $x_0$  is a *critical point*, i.e.,  $\mathbb{D}f(x_0) = 0$ .

**Remark 6.9** This is only a necessary condition but not sufficient. For example, in  $\mathbb{R}^1$ ,  $f(x) = x^2$  at  $(0, 0)$  or in  $\mathbb{R}^2$ ,  $f(x, y) = x^2 - y^2$  at  $(0, 0)$ .



For a critical point that is not an extreme point, we call it a saddle point.

**Proof 1.** (Sketch).

Assume  $\mathbb{D}f(x_0) \neq 0$ . Then, WLOG,  $\exists v \in \mathbb{R}^n$  s.t.  $\mathbb{D}f(x_0)(v) = c > 0$ . By definition of differential, choose  $\delta > 0$  s.t.

$$\|f(x_0 + h) - f(x_0) - \underbrace{\mathbb{D}f(x_0)(h)}_{=\epsilon}\| < \frac{c}{2\|v\|} \cdot \|h\| \quad \forall \|h\| < \delta.$$

Choose  $h = \lambda v$  with  $\lambda > 0$  and  $\|h\| < \delta$ . Then, by triangle inequality,

$$f(x_0 + \lambda v) - f(x_0) > 0 \quad \text{but} \quad f(x_0 - \lambda v) - f(x_0) < 0.$$

Contradiction!

Q.E.D. ■

**Definition 6.7.4 (Positive/Negative (Semi)definite).** A bilinear form  $B : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$  is call *positive definite* (or *negative definite*) if  $B(x, x) > 0$  (or  $< 0$ )  $\forall x \in \mathbb{R}^n$ ,  $x \neq 0$ . We say  $B$  is *positive* (or *negative*) *semidefinite* if  $B(x, x) \geq 0$  (or  $\leq 0$ )  $\forall x \in \mathbb{R}^n$ .

**Definition 6.7.5 (Major Diagonal Factors).** Recall  $B$  is determined by a matrix  $H$  as follows:

$$B(x, x) = xHx^\top, \quad \text{where } H = \begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{bmatrix}_{n \times n}$$

The *major diagonal factors* are given by

$$\begin{aligned} \Delta_1 &= \det \begin{pmatrix} a_{11} \end{pmatrix} = a_{11} \\ \Delta_2 &= \det \begin{pmatrix} a_{11} & a_{21} \\ a_{12} & a_{22} \end{pmatrix} \\ &\vdots \\ \Delta_n &= \det(H). \end{aligned}$$

**Lemma 6.7.6 :**

- $H$  is positive definite  $\iff \Delta_k > 0 \quad \forall k = 1, \dots, n$
- $H$  is positive semi-definite  $\implies \Delta_k \geq 0 \quad \forall k = 1, \dots, n$ .
- $H$  is negative definite  $\iff (-H)$  is positive definite.

**Example 6.7.7**

$$H = \begin{bmatrix} 2 & -1 \\ -1 & 3 \end{bmatrix} \implies \Delta_1 = 2, \Delta_2 = 5 \implies H \text{ is positive definite.}$$

**Theorem 6.7.8 Second Order Sufficient Condition**

Suppose  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  is of class  $\mathcal{C}^2$  and  $x_0 \in A$  is a critical point (i.e.,  $\mathbb{D}f(x) = 0$ ).

- If  $H_f(x_0)$  is negative (or positive) definite, then  $x_0$  is a local maximum (or minimum).
- If  $x_0$  is a local maximum (or minimum), then  $H_f(x_0)$  is negative (or positive) semidefinite.

**Remark 6.10**

- $\text{Max of } f \iff \text{Min of } (-f)$
- *About minimum point:*
  - $\Delta_k > 0 \quad \forall k, H_f(x_0) \text{ is positive definite} \implies x_0 \text{ is local minimum.}$
  - $x_0 \text{ is a local minimum} \implies H_f(x_0) \text{ is positive semidefinite} \implies \Delta_k \geq 0 \quad \forall k.$
  - $\Delta_k < 0 \text{ for some } k \implies x_0 \text{ is not a local minimum.}$

- *About maximum point:*

- $\Delta_k < 0$  for odd  $k$  and  $\Delta_k > 0$  for even  $k \implies (-H_f(x_0))$  is negative definite  $\implies H_f(x_0)$  is negative definite  $\implies x_0$  is local maximum.
- $x_0$  is local maximum  $\implies H_f(x_0)$  is negative semidefinite  $\implies \Delta_k \leq 0$  for odd  $k$  and  $\Delta_k \geq 0$  for even  $k$ .
- $\Delta_k < 0$  for some even  $k \implies x_0$  is not a local maximum  $\implies x_0$  is a saddle point.

**Proof2.** (of ①)

- Set-up: Suppose  $H_f$  is negative definite. Need to show:

$$\exists \delta > 0 \text{ s.t. } \|y - x\| < \delta \implies f(y) \leq f(x_0). \quad (\star)$$

**Scartch:**

By Taylor's Theorem

$$\begin{aligned} f(y) &= f(x_0) + \underbrace{\mathbb{D}f(x_0)}_{=0, \text{critical point}} (y - x_0) + \frac{1}{2} \mathbb{D}^2 f(c)(y - x_0, y - x_0) \\ f(y) - f(x_0) &= \frac{1}{2} \mathbb{D}^2 f(c)(y - x_0, y - x_0). \end{aligned}$$

If  $\mathbb{D}^2 f(c)$  is negative semidefinite, we are done with the proof. However, we only know definiteness at  $x_0$ . Let's add and subtract  $\mathbb{D}^2 f(x_0)$ :

$$f(y) - f(x_0) = \frac{1}{2} \underbrace{\mathbb{D}^2 f(x_0)(y - x_0, y - x_0)}_{\text{negative}} + \frac{1}{2} \underbrace{\left[ \mathbb{D}^2 f(c) - \mathbb{D}^2 f(x_0) \right]}_{\text{make it small}} (y - x_0, y - x_0)$$

- Consider the function

$$g(x) = \mathbb{D}^2 f(x_0)(x, x) : \mathbb{R}^n \rightarrow \mathbb{R}.$$

Denote  $\mathbb{D}^2 f(x_0) = H$ , then  $g(x) = H(x, x)$ .  $g$  is continuous. Then,  $\exists \bar{x} \in S = \{x \in \mathbb{R}^n \mid \|x\| = 1\}$  s.t.

$$H(x, x) \leq H(\bar{x}, \bar{x}).$$

*Extreme Value Theorem:* Continuous function on closed and bounded set attains its maximum and minimum. Since  $H$  is negative definite,  $H(\bar{x}, \bar{x}) < 0$ . Let  $\varepsilon = -H(\bar{x}, \bar{x}) > 0$ . Then, for any  $h \in \mathbb{R}^n$  with  $h \neq 0$ , we have

$$H(h, h) = \|h\|^2 \cdot H\left(\frac{h}{\|h\|}, \frac{h}{\|h\|}\right) \leq \|h\|^2 \cdot H(\delta x, \bar{x}).$$



So,

$$H(h, h) \leq -\varepsilon \|h^2\| \quad (\text{I})$$

- Prove  $(\star)$  is true in a neighborhood.

By continuity of  $\mathbb{D}^2 f$  at  $x_0$ ,  $\exists \delta > 0$  s.t.

$$\|y - x_0\| < \delta \implies y \in A, \underbrace{\|\mathbb{D}^2 f(y) - \mathbb{D}^2 f(x_0)\|}_{\text{operator norm}} < \frac{\varepsilon}{2} \quad (\text{II})$$

*Operator norm satisfies:*  $\|T(x, y)\| \leq \|T\| \cdot \|x\| \cdot \|y\|$ .

By Taylor's Formula, because  $\mathbb{D}f(x_0) = 0$ , we have

$$f(y) - f(x) = \frac{1}{2} \mathbb{D}^2 f(c)(h, h),$$

where  $y \in B(x_0, \delta)$ ,  $h = y - x_0$ , and  $c \in [x_0, y]$ . Note that

$$\begin{aligned} \mathbb{D}^2 f(c)(h, h) &= [\mathbb{D}^2 f(c) - \mathbb{D}^2 f(x_0)](h, h) + \mathbb{D}^2 f(x_0)(h, h) \\ &\leq \|\mathbb{D}^2 f(c) - \mathbb{D}^2 f(x_0)\| \cdot \|h\|^2 + (-\varepsilon) \|h\|^2 && \text{By (I)} \\ &\leq \frac{1}{2} \varepsilon \|h\|^2 + (-\varepsilon) \|h\|^2 && \text{By (II)} \\ &= -\frac{\varepsilon}{2} \|h\|^2 \leq 0. \end{aligned}$$

Then,  $f(y) \leq f(x) \quad \forall y \in B(x_0, \delta)$ . So,  $x_0$  is the local maximum.

Q.E.D. ■

### Example 6.7.9

Find and classify the critical points for  $f(x, y, z) = \cos 2x \sin y + z^2$ .

**Solution 3.**

- Find the critical point:

$$\frac{\partial f}{\partial x} = -2 \sin 2x \sin y; \quad \frac{\partial f}{\partial y} = \cos 2x \cos y; \quad \frac{\partial f}{\partial z} = 2z.$$

Set

$$\frac{\partial f}{\partial x} = \frac{\partial f}{\partial y} = \frac{\partial f}{\partial z} = 0.$$

Then,

$$\begin{cases} -2 \sin 2x \sin y = 0 \\ \cos 2x \cos y = 0 \\ 2z = 0 \end{cases} \implies \begin{cases} x = \frac{k\pi}{2} \\ y = \frac{2j+1}{2}\pi \\ z = 0 \end{cases} \quad \text{or} \quad \begin{cases} x = \frac{2k+1}{4}\pi \\ y = j\pi \\ z = 0. \end{cases}$$

- Classify critical points:

Compute the Hessian

$$\frac{\partial^2 f}{\partial x^2} = -4 \cos 2x \sin y; \quad \frac{\partial^2 f}{\partial y \partial x} = -2 \sin 2x \cos y; \quad \frac{\partial^2 f}{\partial z \partial x} = 0$$

$$\frac{\partial^2 f}{\partial y^2} = -\cos 2x \sin y; \quad \frac{\partial^2 f}{\partial z \partial y} = 0; \quad \frac{\partial^2 f}{\partial z^2} = 2.$$

So,

$$H_f(x) = \begin{bmatrix} -4 \cos 2x \sin y & -2 \sin 2x \cos y & 0 \\ -2 \sin 2x \cos y & -\cos 2x \sin y & 0 \\ 0 & 0 & 2 \end{bmatrix}.$$

**Case I**  $x = \frac{k\pi}{2}, y = \frac{2j+1}{2}\pi, z = 0$ . Then,

$$H_f\left(\frac{k\pi}{2}, \frac{2j+1}{2}\pi, 0\right) = \begin{bmatrix} -4(-1)^k(-1)^j & 0 & 0 \\ 0 & -1(-1)^k(-1)^j & 0 \\ 0 & 0 & 2 \end{bmatrix}.$$

Then,  $\Delta_1 = -4(-1)^{j+k}, \Delta_2 = 4(-1)^{2k}(-1)^{2j} = 4 > 0$ , and  $\Delta_3 = 2 \cdot \Delta_2 = 8 > 0$ .

- If  $j + k$  is odd, then  $\Delta_1 > 0$ . Then,  $H_f$  is positive definite, and the critical point is a local minimum.
- If  $j + k$  is even, then  $\Delta_1 < 0$ . Then, the critical point is not a local minimum. But  $\Delta_3 = 0 > 0$ , so it cannot be a local maximum. Hence, it must be a saddle point.

**Case II**  $x = \frac{2k+1}{4}\pi, y = j\pi, z = 0$ . Then,

$$H_f\left(\frac{2k+1}{4}\pi, j\pi, 0\right) = \begin{bmatrix} 0 & (-2)(-1)^k(-1)^j & 0 \\ (-2)(-1)^k(-1)^j & 0 & 0 \\ 0 & 0 & 2 \end{bmatrix}.$$

Then,  $\Delta_1 = 0, \Delta_2 = -(-2)(-1)^{k+j} \cdot (-2)(-1)^{k+j} = -4(-1)^{2(k+j)} = -4 < 0$ , and  $\Delta_3 = 0$ . As  $\Delta_2 < 0$ , they are saddle points.

- Conclusion:

$$\left(\frac{k\pi}{2}, \frac{2j+1}{2}\pi, 0\right) \begin{cases} \text{local minimum when } k+j \text{ is odd} \\ \text{saddle point when } k+j \text{ is even.} \end{cases}$$

$$\left(\frac{2k+1}{4}\pi, j\pi, 0\right) : \text{saddle point.}$$

□

## 7 Inverse and Implicit Function Theorem

### 7.1 Inverse Function Theorem

#### 7.1.1 Linear Case.

- Consider a linear map:  $y = f(x) : \mathbb{R}^n \rightarrow \mathbb{R}^n$  given by

$$\begin{cases} y_1 &= a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \\ &\vdots \\ y_n &= a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n. \end{cases}$$

Or, in matrix notation:

$$Ax = y \quad (\star)$$

- Given  $y \in \mathbb{R}^n$ ,  $(\star)$  is a linear system of equations.
- Fact:**  $(\star)$  has unique solution  $x \iff A$  is invertible. i.e.,  $\det(A) \neq 0$ . In this case, the solution is given by  $x = A^{-1}y$ .
- $x = A^{-1}y$  is the inverse function of  $y = f(x)$ .

#### 7.1.2 When can we solve a nonlinear system?.

- Nonlinear System:

$$\begin{cases} f_1(x_1, x_2, \dots, x_n) &= y_1 \\ &\vdots \\ f_n(x_1, x_2, \dots, x_n) &= y_n \end{cases}, \quad \text{or } f(x) = y.$$

*In order to have inverse, dimension must match.*

- Notation 7.3.**

- $y = f(x) : A \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$ , where  $A$  is open and  $f$  is differentiable on  $A$ . Suppose  $y = (y_1, y_2, \dots, y_n)$ ,  $x = (x_1, x_2, \dots, x_n)$ , and  $f = (f_1, f_2, \dots, f_n)$ .
- $\mathbb{D}f(x) = \left( \frac{\partial f_j}{\partial x_i} \right)_{ij}$  and  $J_f(x) = \det(\mathbb{D}f(x))$  is the *Jacobian determinant* of  $f$  at  $x$ .

#### Theorem 7.1.4 Inverse Function Theorem

Let  $y = f(x) : A \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$  be of class  $\mathcal{C}^1$ . Suppose  $x_0 \in A$  and  $J_f(x_0) \neq 0$ . Then,  $\exists$  neighborhoods  $U$  of  $x_0$  and  $W$  of  $y_0 = f(x_0)$  s.t.

- $f(U) = W$  and  $f : U \rightarrow W$  has an inverse  $f^{-1} : W \rightarrow U$
- $f^{-1} : W \rightarrow U$  is of class  $\mathcal{C}^1$ . Additionally, if  $f \in \mathcal{C}^r$ , then  $f^{-1} \in \mathcal{C}^r$ .
- $\mathbb{D}f^{-1}(y) = [\mathbb{D}f(x)]^{-1} \quad \forall y \in W \text{ and } y = f(x)$ .

► **Proof 1 of Inverse Function Theorem**

**Theorem (Contraction Mapping Principle / CMP)** Let  $\mathcal{X}$  be a complete metric space and  $\varphi : \mathcal{X} \rightarrow \mathcal{X}$ . Suppose  $\exists 0 < k < 1$  s.t.

$$d(\varphi(x), \varphi(y)) \leq k \cdot d(x, y) \quad \forall x, y \in \mathcal{X}.$$

Then,  $\exists$  unique fixed point  $x^*$  s.t.  $\varphi(x^*) = x^*$ .

**Step 1 Reductions**

- We may assume that  $\mathbb{D}f(x_0) = I$ .

In fact, let  $T = \mathbb{D}f(x_0)$ . Then,  $J_f(x_0) \neq 0 \implies T^{-1}$  exists. Consider a new map:  $T^{-1} \circ f : A \rightarrow \mathbb{R}$ . Then,

$$\begin{aligned} \mathbb{D}(T^{-1} \circ f) &= \mathbb{D}T^{-1}(f(x_0)) \circ \mathbb{D}f(x_0) \\ &= T^{-1} \circ T \\ &= I. \end{aligned}$$

If the inverse of  $T^{-1} \circ f$  exists, then the inverse of  $f$  also exists. So, once the identity case is true, we just multiply  $T^{-1}$  to  $f$  and we can get the general case is true.

- We may assume that  $x_0 = 0$  and  $f(x_0) = 0$ .

To see this, let  $h(x) = f(x + x_0) - f(x_0)$ . Then,  $h(0) = 0$  and  $\mathbb{D}h(0) = \mathbb{D}f(x_0)$ . If the inverse of  $h(x)$  exists, then the equation  $f(x) = y$  can be solved:

$$\begin{aligned} f(x) &= h(x - x_0) + f(x_0) = y \\ h(x - x_0) &= y - f(x_0) \\ x - x_0 &= h^{-1}(y - f(x_0)) \\ x &= h^{-1}(y - f(x_0)) + x_0. \end{aligned}$$

**Step 2 Existence of Inverse**

- By reduction above, we have  $x_0 = 0$ ,  $y_0 = f(x_0) = 0$ ,  $\mathbb{D}f(x_0) = \mathbb{D}f(0) = I$ .

WTS:  $\exists$  neighborhoods  $U, W$  of 0 s.t. the map  $y = f(x) : U \rightarrow W$  has an inverse in  $W$ . i.e.,  $\forall y \in W, \exists$  unique  $x \in U$  s.t.  $y = f(x)$ .

For a fixed  $y \in \mathbb{R}^n$ , define  $g_y(x) := y + x - f(x) : A \rightarrow \mathbb{R}^n$ .

If  $g_y(x)$  has a fixed point:  $g_y(x^*) = x^* = y + x^* - f(x^*) \implies y - f(x^*) = 0$ . So, we want to show  $g_y(x)$  has a unique fixed point.

- Construction of neighborhoods  $U$  and  $W$ .

Let  $g(x) = x - f(x)$ . Then,

$$\mathbb{D}g(0) = I - \mathbb{D}f(0) = I - I = 0.$$

Since  $f \in \mathcal{C}^1$ ,  $g \in \mathcal{C}^1$ . Then,  $\mathbb{D}g(x)$  is continuous at 0. Then,  $\forall \varepsilon = \frac{1}{2n}$ ,  $\exists \delta > 0$  s.t.

$$\|x - 0\| < \delta \implies \|\mathbb{D}g_i(x) - \mathbb{D}g_i(0)\| = \|\mathbb{D}g_i(x) - 0\| = \|\mathbb{D}g_i(x)\| < \frac{1}{2n},$$

where  $g = (g_1, g_2, \dots, g_n)$ .

Apply MVT to each of  $g_i$ , we obtain  $\forall x \in \overline{B}(x_0, \delta)$ ,  $\exists c_i \in [0, x]$  s.t.

$$g_i(x) = g_i(x) - g_i(0) = \mathbb{D}g_i(c_i)(x - 0).$$

So,

$$\begin{aligned} \|g(x)\| &\leq \sum_{i=1}^n \|g_i(x)\| = \sum_{i=1}^n |\mathbb{D}g_i(c_i) \cdot x| \\ &\leq \sum_{i=1}^n \|\mathbb{D}g_i(c_i)\| \cdot \|x\| && \text{[operator norm]} \\ &\leq \sum_{i=1}^n \frac{1}{2n} \|x\| && \text{[continuity of } \mathbb{D}g\text{]} \\ &= \frac{1}{2} \|x\|. \end{aligned}$$

i.e.,  $\|g(x)\| \leq \frac{1}{2} \|x\|$ . Thus,  $g : \overline{B}(0, \delta) \rightarrow \overline{B}(0, \frac{1}{2}\delta) \subset \overline{B}(0, \delta)$  is a contraction map. Let  $W = B(0, \frac{\delta}{2})$  and  $U = \{x \in B(0, \delta) : f(x) \in W\}$ . **WTS:  $U$  and  $W$  are the desired neighborhoods.**

- Show existence of  $f^{-1} : W \rightarrow U$ .

Fix  $y \in W$ . Then,  $\forall x \in \overline{B}(0, \delta)$ ,

$$\begin{aligned} \|g_y(x)\| &= \|y + g(x)\| \leq \|y\| + \|g(x)\| \\ &< \frac{\delta}{2} + \frac{1}{2}\delta = \delta \quad \left[ y \in W = B\left(0, \frac{\delta}{2}\right), \|g(x)\| \leq \frac{1}{2}\|x\|, x \in U = B(0, \delta) \right] \end{aligned}$$

Then,  $g_y(x) : \overline{B}(0, \delta) \rightarrow \overline{B}(0, \delta)$  and  $g_y$  is also a contraction map with  $k = \frac{1}{2}$ . Then, by CMP,  $\exists$  unique  $x$  s.t.  $g_y(x) = x$ . Then,

$$\begin{aligned} g_y(x) &= y + x - f(x) = x \\ y - f(x) &= 0 \implies y = f(x). \end{aligned}$$

So, for fixed  $y$ ,  $\exists$  unique  $x$  s.t.  $y = f(x)$ . Then,  $f$  is a bijection, and thus the inverse exists.

**Step 3** Continuity of  $f^{-1}$ .

**WTS:  $f^{-1}$  is Lipschitz continuous.**

Fix  $y_1, y_2 \in W$ . Let  $x_i = f^{-1}(y_i)$  for  $i = 1, 2$ . Then,

$$\begin{aligned} \|f^{-1}(y_1) - f^{-1}(y_2)\| &= \|x_1 - x_2\| = \|g(x_1) + f(x_1) - g(x_2) - f(x_2)\| \\ &\leq \|g(x_1) - g(x_2)\| + \|f(x_1) - f(x_2)\| \\ &= \|g(x_1) - g(x_2)\| + \|y_1 - y_2\|. \end{aligned}$$

Since  $\|\mathbb{D}g(x)\| \leq \frac{1}{2}$  for  $x \in \overline{B}(0, \delta)$ , by Mean Value Inequality,

$$\|g(x_1) - g(x_2)\| \leq \frac{1}{2}\|x_1 - x_2\|.$$

Then,

$$\|x_1 - x_2\| \leq \frac{1}{2}\|x_1 - x_2\| + \|y_1 - y_2\|.$$

So,

$$\frac{1}{2}\|x_1 - x_2\| \leq \|y_1 - y_2\| \implies \|x_1 - x_2\| \leq 2\|y_1 - y_2\|.$$

That is,

$$\|f^{-1}(y_1) - f^{-1}(y_2)\| \leq 2\|y_1 - y_2\| \quad (\star)$$

Thus,  $f^{-1}$  is Lipschitz and thus continuous.

#### Step 4 Differentiability of $f^{-1}$

- **Proposition**  $[\mathbb{D}f(0)]^{-1}$  exists and  $\mathbb{D}f(x)$  is continuous at 0  $\implies \exists \delta' > 0$  s.t.  $[\mathbb{D}f(x)]^{-1}$  exists and bounded by  $M$ :

$$\underbrace{\|\mathbb{D}f(x) \cdot (v)\|}_{\text{operator norm}} \leq \|M\| \cdot \|v\| \quad \forall \|x\| < \delta' \text{ and } v \in \mathbb{R}^n.$$

- **WTS:**  $f^{-1}(y)$  is differentiable at any fixed point  $y_0 \in W$  and

$$\mathbb{D}f^{-1}(y_0) = [\mathbb{D}f(x_0)]^{-1} \quad \text{with } y_0 = f(x_0).$$

Fix  $y_0 \in W$ . Then,

$$\begin{aligned} & \frac{\|f^{-1}(y) - f^{-1}(y_0) - \mathbb{D}f^{-1}(y_0) \cdot (y - y_0)\|}{\|y - y_0\|} \\ &= \frac{\|[\mathbb{D}f(x_0)]^{-1} \cdot [\mathbb{D}f(x_0) \cdot f^{-1}(y) - \mathbb{D}f(x_0) \cdot f^{-1}(y_0) - (y - y_0)]\|}{\|y - y_0\|} \quad [\text{factor out } \mathbb{D}f^{-1}(y_0) = [\mathbb{D}f(x_0)]^{-1}] \\ &= \frac{\|[\mathbb{D}f(x_0)]^{-1} \cdot [\mathbb{D}f(x_0)(x - x_0) - (f(x) - f(x_0))]\|}{\|f(x) - f(x_0)\|} \quad [y = f(x)] \\ &= \frac{\|[\mathbb{D}f(x_0)]^{-1} \cdot [\mathbb{D}f(x_0)(x - x_0) - (f(x) - f(x_0))]\| \cdot \|x - x_0\|}{\|f(x) - f(x_0)\| \cdot \|x - x_0\|} \quad [\text{Multiply by magic 1}] \\ &\leq \frac{2\|[\mathbb{D}f(x_0)]^{-1}[\mathbb{D}f(x_0)(x - x_0) - (f(x) - f(x_0))]\|}{\|x - x_0\|} \quad [\text{Lipschitz continuity, Eq } (\star)] \\ &\rightarrow 0 \quad \text{as } x \rightarrow x_0. \end{aligned}$$

So,  $f^{-1}$  is differentiable, and

$$[\mathbb{D}f^{-1}(y)] = [\mathbb{D}f(x)]^{-1}.$$

Q.E.D. ■

### Example 7.1.5

Investigate the invertibility (both local and global) for the map  $W = (u, v) = f(x, y) : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  given by  $u = e^x \cos y$  and  $v = e^x \sin y$ .

#### Solution 2.

Firstly, we know  $f \in C^\infty$ . Compute the Jacobian determinant:

$$\begin{aligned} J_f(x, y) &= \det(\mathbb{D}f(x)) = \det \begin{pmatrix} \partial u / \partial x & \partial u / \partial y \\ \partial v / \partial x & \partial v / \partial y \end{pmatrix} \\ &= \det \begin{pmatrix} e^x \cos y & -e^x \sin y \\ e^x \sin y & e^x \cos y \end{pmatrix} \\ &= e^{2x} \cos^2 y + e^{2x} \sin^2 y \\ &= e^{2x} > 1. \end{aligned}$$

So, by the Inverse Function Theorem,  $f$  is invertible locally at any point, and the differentiable of the inverse is given by

$$\mathbb{D}f^{-1}(u, v) = [\mathbb{D}f(x, y)]^{-1} = \begin{bmatrix} e^x \cos y & -e^x \sin y \\ e^x \sin y & e^x \cos y \end{bmatrix}^{-1}.$$

Now, let's examine if  $f$  is globally invertible (i.e., if  $f$  is a one-to-one function on  $\mathbb{R}^2$ ). Note that

$$f(x_0, y_0) = e^{x_0} \cos y_0$$

and

$$f(x_0, y_0 + 2\pi) = e^{x_0} \cos(y_0 + 2\pi) = e^{x_0} \cos(y_0) \text{ and } f(x_0, y_0 - 2\pi) = e^{x_0} \cos(y_0 - 2\pi) = e^{x_0} \cos(y_0).$$

So,  $f$  is not globally invertible since  $f$  is not an injection. □

**Remark 7.1**  $f$  can be written in complex notation:  $f(z) = e^z$ , where  $z = x + iy \in \mathbb{C}$ . Then,

$$f(z) = e^z = e^{x+iy} = e^x (\cos x + i \sin y).$$

Meanwhile,  $f^{-1}(z) = \ln(z)$ .

## 7.2 Implicit Function Thm and Applications

### Motivation

- Given a function  $f : \mathbb{R} \rightarrow \mathbb{R}$ . Consider an equation  $f(y) = x$ . If it can be solved for  $y$  (uniquely in terms of  $x$ ), then the solution  $y = g(x)$  is the inverse of  $f$ . That is,  $(f \circ g)(x) = x$ .
- Reinterpretation of Inverse:  
Rewrite  $f(y) = x$  as  $x - f(y) = 0$  ①.  
Then,  $f$  is invertible  $\iff$  Equation ① is solvable for  $y$ .
- **Question:** When can we solve a general equation for  $y$ ,  $F(x, y) = 0$  ( $F : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^m$ )?  
The solution of  $F(x, y) = 0$ , denoted by  $y = g(x)$ , is called the *implicit function* determined by  $F(x, y) = 0$ .

#### Example 7.2.1

Consider equation  $x^2 + y^2 - 1 = 0$  to be  $F(x, y) : \mathbb{R}^1 \times \mathbb{R}^1 \rightarrow \mathbb{R}^1$ .

Given  $(x_0, y_0)$  s.t.  $F(x_0, y_0) = 0$  with  $y_0 \neq 0$ . Then,  $\exists$  a unique solution

$$y = \begin{cases} \sqrt{1 - x^2} & \text{if } y_0 > 0 \\ -\sqrt{1 - x^2} & \text{if } y_0 < 0. \end{cases}$$

in the neighborhood of  $x_0$ .

Note that  $\left. \frac{\partial F}{\partial y} \right|_{y=y_0} = 2y_0 \neq 0$  when  $y_0 \neq 0$ .

#### Theorem 7.2.2 Implicit Function Theorem

Let  $A \subset \mathbb{R}^n \times \mathbb{R}^m$  and  $F(x, y) : A \rightarrow \mathbb{R}^m$  be of class  $\mathcal{C}^1$ . Suppose  $(x_0, y_0) \in A$  with  $F(x_0, y_0) = 0$ . If

$$\begin{aligned} \Delta = \det \left( \frac{\partial F}{\partial y} \right) &= \det \left( \frac{\partial(F_1, \dots, F_m)}{\partial(y_1, \dots, y_m)} \right) \\ &= \det \begin{bmatrix} \frac{\partial F_1}{\partial y_1} & \dots & \frac{\partial F_1}{\partial y_m} \\ \vdots & \ddots & \vdots \\ \frac{\partial F_m}{\partial y_1} & \dots & \frac{\partial F_m}{\partial y_m} \end{bmatrix} \neq 0 \quad \text{at } (x_0, y_0), \end{aligned}$$

then  $\exists$  neighborhoods  $U$  of  $x_0$ ,  $V$  of  $y_0$ , and a unique function  $y = f(x) : U \rightarrow V$  such that  $F(x, f(x)) = 0 \quad \forall x \in U$ . i.e.,  $y = f(x)$  is the solution of  $F(x, y) = 0$ .

Furthermore, if  $F \in \mathcal{C}^r$ , then  $f \in \mathcal{C}^r$ .

#### Remark 7.2

- $y = f(x)$  is called the *implicit function determined by the equation*  $F(x, y) = 0$  *based at the point*  $(x_0, y_0)$ .



- *Differential of implicit function:*

Suppose  $n = m = 1$  and  $F(x, y) = 0$ . Then, by chain rule,

$$\begin{aligned} \frac{\partial F}{\partial x} \cdot \frac{dx}{dx} + \frac{\partial F}{\partial y} \cdot \frac{dy}{dx} &= 0 \\ \frac{dy}{dx} &= -\frac{\partial F / \partial x}{\partial F / \partial y}. \end{aligned}$$

In the general case, let  $y = f(x) = (f_1, \dots, f_m) : \mathbb{R}^n \rightarrow \mathbb{R}^m$ . Let  $f$  be the implicit function determined by  $F(x, y) = 0$ . Then,

$$\mathbb{D}f = -\left(\frac{\partial F}{\partial y}\right)^{-1} \cdot \left(\frac{\partial F}{\partial x}\right).$$

### ► Proof 1 of Implicit Function Theorem

Given  $F(x, y) = 0$ ,  $A \subset \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^m$ . Consider the map  $G : A \subset \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n \times \mathbb{R}^m$  given by

$$G(x, y) = (x, F(x, y)).$$

We want to use Inverse Function Theorem. So, we need a map that maps to the same dimension.

Suppose  $G^{-1}$  exists in a neighborhood of  $(x_0, y_0)$ . Write

$$G^{-1}(x, 0) = (x, f(x)).$$

Then,  $y = f(x)$  is the solution of  $F(x, y) = 0$  because

$$\begin{aligned} G(x, f(x)) &= (x, 0) \\ &= (x, F(x, f(x))). \end{aligned}$$

So,  $F(x, f(x)) = 0$ .

It remains to show that  $G$  is invertible. This follows from the inverse function theorem. Consider

$$\mathbb{D}G \Big|_{(x,y)=(x_0,y_0)} = \left[ \begin{array}{cccc|cccc} 1 & 0 & \cdots & 0 & 0 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 & 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 & 0 & 0 & \cdots & 0 \\ \hline \frac{\partial F_1}{\partial x_1} & \cdots & \frac{\partial F_1}{\partial x_n} & \frac{\partial F_1}{\partial y_1} & \cdots & \frac{\partial F_1}{\partial y_m} & \cdots & \frac{\partial F_1}{\partial y_m} \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots & \cdots & \vdots \\ \frac{\partial F_m}{\partial x_1} & \cdots & \frac{\partial F_m}{\partial x_n} & \frac{\partial F_m}{\partial y_1} & \cdots & \frac{\partial F_m}{\partial y_m} & \cdots & \frac{\partial F_m}{\partial y_m} \end{array} \right].$$

So,

$$J_G(x_0, y_0) = \det \begin{bmatrix} \partial F_1 / \partial y_1 & \cdots & \partial F_1 / \partial y_m \\ \vdots & \ddots & \vdots \\ \partial F_m / \partial y_1 & \cdots & \partial F_m / \partial y_m \end{bmatrix} = \Delta \neq 0,$$

as assumed in implicit function theorem. Therefore, by the inverse function theorem,  $G$  is invertible.

Q.E.D. ■

### Example 7.2.3

Discuss the solvability of  $\begin{cases} y + x + uv = 0 \\ uxy + v = 0 \end{cases}$  for  $u, v$  in terms of  $x, y$  near the point  $(0, 0, 0, 0)$  and the point  $(1, 1, \sqrt{2}, -\sqrt{2})$ . If impossible, compute  $\frac{\partial u}{\partial x}$  and  $\frac{\partial v}{\partial x}$  if exists.

**Solution 2.**

$$F(x, y, u, v) = 0 \text{ and } \begin{cases} F_1 = y + x + uv \\ F_2 = uxy + v. \end{cases} \text{ Let's compute } \Delta:$$

$$\begin{aligned} \Delta &= \det \left( \frac{\partial(F_1, F_2)}{\partial(u, v)} \right) = \det \begin{bmatrix} \partial F_1 / \partial u & \partial F_1 / \partial v \\ \partial F_2 / \partial u & \partial F_2 / \partial v \end{bmatrix} \\ &= \det \begin{bmatrix} v & u \\ xy & 1 \end{bmatrix} \\ &= v - uxy. \end{aligned}$$

Then,  $\Delta(0, 0, 0, 0) = 0$ . So, Implicit Function Theorem does not apply. On the other hand,

$$\Delta(1, 1, \sqrt{2}, -\sqrt{2}) = -\sqrt{2} - \sqrt{2} = -2\sqrt{2} \neq 0.$$

So, by Implicit Function Theorem,  $\exists$  unique solution  $u = u(x, y)$  and  $v = v(x, y)$  in a neighborhood.

Furthermore, the differentiable is given by

$$\begin{aligned} \frac{\partial(u, v)}{\partial(x, y)} &= - \left( \frac{\partial F}{\partial(u, v)} \right)^{-1} \left( \frac{\partial F}{\partial(x, y)} \right) \\ &= - \begin{bmatrix} v & u \\ xy & 1 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 1 \\ uy & ux \end{bmatrix}. \end{aligned}$$

□

**Theorem 7.2.4 Application: Domain-Straightening Theorem**

Let  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$ . Suppose  $\mathbb{D}f(x_0) \neq 0$  and  $f(x_0) = 0$ . Then,  $\exists$  open sets  $U$  and  $V$  (with  $x_0 \in V$ ) and invertible map  $h : U \rightarrow V$  s.t.  $f(h(x_1, \dots, x_n)) = x_0$ .

**Remark 7.3** Under change of variables  $h$ , one can flatten the level curves of function  $f(x)$ .

**Theorem 7.2.5 Application: Range-Straightening Theorem**

Suppose  $f : A \subset \mathbb{R}^p \rightarrow \mathbb{R}^n$  with  $p < n$  and rank of  $\mathbb{D}f(x_0) = p$ . Then,  $\exists$  neighborhoods  $U$ ,  $V$ , and invertible map  $g : U \rightarrow V$  s.t.  $g \circ f(x_1, \dots, x_p) = (x_1, \dots, x_p, 0, \dots, 0)$ .

**7.3 Constrained Extrema****7.3.1 Morse Theory: Local Behavior Near a Critical Point**

Let  $f(x) : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be of class  $\mathcal{C}^2$  and  $x_0$  is a critical point. Then, one can use  $H_f(x_0)$  to classify critical point  $x_0$ .

- Morse Theory makes this classification more precise.
- **Lemma 7.3.1 Morse Lemma:** Let  $f(x) : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be of class  $\mathcal{C}^2$  with critical point  $x_0 \in A$ . If  $H_f(x_0)$  is nondegenerate (i.e.,  $\det(H_f(x_0)) \neq 0$ ), then  $\exists$  neighborhoods  $U$  for  $x_0$  and  $V$  for 0, and invertible map  $g : V \rightarrow U$  s.t. the function  $h = f \circ g$  has the form

$$h(y) = f(x_0) - [y_1^2 + \dots + y_\lambda^2] + [y_{\lambda+1}^2 + \dots + y_n^2],$$

where  $\lambda$  is an integer called the *index* of  $f$  at  $x_0$ .

- Interpretation/Application:
  1.  $\lambda = 0$ :  $x_0$  is a local minimum. Paraboloid open up.
  2.  $\lambda = n$ :  $x_0$  is a local maximum. Paraboloid open down.
  3.  $0 < \lambda < n$ :  $x_0$  is a saddle point. Hyperboloid.
- What is  $\lambda$ ?  
 $\lambda$  (the index of  $f$  at  $x_0$ ) is the number of negative eigenvalues of  $H_f(x_0)$ .

**Example 7.3.2**

Determine the shape of the surface given by  $z = x^2 + 3xy - y^2$  near critical point  $(0, 0)$ .

**Solution 1.**

$\mathbb{D}f = \begin{pmatrix} 2x + 3y & 3x - 2y \end{pmatrix}$ . Therefore,

$$H_f(x, y) = \begin{bmatrix} 2 & 3 \\ 3 & -2 \end{bmatrix}$$

The eigenvalues are  $t = \pm\sqrt{13}$ . So, index  $\lambda = 1$ . As  $0 < \lambda < n$ ,  $(0, 0)$  is a saddle point. The shape is thus a hyperboloid. □

### 7.3.2 Constrained Extremal Problem

**Goal:** To maximum (or minimize) a function  $f(x) : \mathbb{R}^n \rightarrow \mathbb{R}$  under the constraint  $g(x) = c$ .

**Tool:** Lagrange Multiplier Method.

#### Theorem 7.3.3 Necessary Condition

Let  $f, g : U \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be of class  $\mathcal{C}^1$ . Assume  $g(x_0) = c_0$  with  $\nabla g(x_0) \neq 0$ . If  $f$  restricted to the surface  $S : g(x) = c_0$  has maximum or minimum at  $x_0$ , then  $\exists \lambda \in \mathbb{R}$  s.t.

$$\nabla f(x_0) = \lambda \nabla g(x_0).$$

**Remark 7.4 (Geometric Meaning)**  $\nabla f(x_0)$  is parallel to  $\nabla g(x_0)$ .

**Proof2.**

- Geometric proof: **WTS:**  $\nabla f(x_0) \perp S$ .

Fix curve  $c(t)$  at  $t_0$ . So,  $c(t_0) = x_0$ . **WTS:**  $\nabla f(x) \perp c'(t)$ .

Since  $f$  restricted to  $S$  has a maximum at  $x_0$ ,  $h(t) = f(c(t))$  has a maximum at  $t_0$ . Then,

$$0 = h'(t_0) = \nabla f(x_0) \cdot c'(t_0) = \langle \nabla f(x_0), c'(t_0) \rangle.$$

So,  $\nabla f(x_0) \perp c'(t_0)$ , and thus  $\nabla f(x_0) \perp S$ .

- Analytical proof: **Substitute the condition**  $g(x) = c_0$  **into**  $f(x)$

Since

$$\nabla g(x_0) = \left( \frac{\partial g}{\partial x_1}, \dots, \frac{\partial g}{\partial x_n} \right) \neq \vec{0},$$

then  $\exists \frac{\partial g}{\partial x_i} \neq 0$  for some  $i = 1, \dots, n$ . WLOG, assume  $\frac{\partial g}{\partial x_n} \neq 0$ . By Implicit Function Theorem, the equation

$$g(x_1, \dots, x_n) = c_0$$

can be uniquely solve for  $x_n$ :

$$x_n = h(x_1, \dots, x_{n-1}).$$

Let  $k(x_1, \dots, x_{n-1}) = f(x_1, \dots, x_{n-1}, h(x_1, \dots, x_{n-1}))$ . Then, the maximum of  $f$  correspond to maximum of  $k$ . Then,

$$0 = \frac{\partial k}{\partial x_i} = \frac{\partial f}{\partial x_i} + \frac{\partial f}{\partial x_n} \cdot \frac{\partial h}{\partial x_i} \quad \text{for } i = 1, \dots, n-1. \quad (1)$$

Furthermore,  $g(x) = c_0$ . So,  $g(x_1, \dots, x_{n-1}, h(x_1, \dots, x_{n-1})) = c_0$ . Then,

$$\frac{\partial g}{\partial x_i} + \frac{\partial g}{\partial x_n} \cdot \frac{\partial h}{\partial x_i} = 0 \quad \text{for } i = 1, \dots, n-1.$$

Then,

$$\frac{\partial h}{\partial x_i} = -\frac{\partial g/\partial x_i}{\partial g/\partial x_n} \quad (2)$$

Substitute (2) into (1):

$$\begin{aligned} \frac{\partial f}{\partial x_i} &= -\frac{\partial f}{\partial x_n} \cdot \frac{\partial h}{\partial x_i} = -\frac{\partial f}{\partial x_n} \cdot \frac{-\partial g/\partial x_i}{\partial g/\partial x_n} \\ &= \underbrace{\frac{\partial f/\partial x_n}{\partial g/\partial x_n}}_{\lambda} \cdot \frac{\partial g}{\partial x_i} \\ &= \lambda \frac{\partial g}{\partial x_i}. \end{aligned}$$

So,

$$\frac{\partial f}{\partial x_i} = \lambda \frac{\partial g}{\partial x_i} \quad \forall i = 1, \dots, n.$$

That is,

$$\nabla f(x) = \lambda \nabla g(x).$$

Q.E.D. ■

#### Theorem 7.3.4 General Procedure to Solve an Extremal Problem

- Solve the equations for  $x \in \mathbb{R}^n$  and  $\lambda \in \mathbb{R}$ :

$$\begin{cases} g(x) = c_0 \\ \nabla f(x) = \lambda \nabla g(x) \end{cases}$$

- Compare values of  $f$  at these points.

#### Example 7.3.5

Find extrema for the function  $f(x, y) = x^2 - y^2$  subject to the constraint  $x^2 + y^2 = 1$ .

**Solution 3.**

Solve the equations:

$$\begin{cases} g(x) = c_0 \\ \nabla f(x) = \lambda \nabla g(x) \end{cases} \implies \begin{cases} x^2 + y^2 = 1 \\ \begin{bmatrix} 2x \\ -2y \end{bmatrix} = \lambda \begin{bmatrix} 2x \\ 2y \end{bmatrix} \end{cases} \implies \begin{cases} x^2 + y^2 = 1 \\ 2x = \lambda 2x \\ -2y = \lambda 2y. \end{cases}$$

- If  $x = 0$ ,  $y = \pm 1$ , and  $\lambda = -1$ .
- If  $y = 0$ ,  $x = \pm 1$ , and  $\lambda = 1$ .

Possible candidates:  $(0, 1)$ ,  $(0, -1)$ ,  $(1, 0)$ , and  $(-1, 0)$ .

- At  $(0, 1)$ ,  $f(0, 1) = 0^2 - 1^2 = -1$ .
- At  $(0, -1)$ ,  $f(0, -1) = 0^2 - (-1)^2 = -1$ .
- At  $(1, 0)$ ,  $f(1, 0) = 1^2 - 0^2 = 1$ .
- At  $(-1, 0)$ ,  $f(-1, 0) = (-1)^2 - 0 = 1$ .

Then,  $(0, 1)$  and  $(0, -1)$  are local minimum, and  $(1, 0)$  and  $(-1, 0)$  are local maximum.

□

### Theorem 7.3.6 Extremal Problem with Multiple Constraints

Maximize/Minimize  $f(x)$  with constraints  $g_1(x) = c_1, \dots, g_m(x) = c_m$ . Then, we solve

$$\begin{cases} g_1(x) = c_1 \\ \vdots \\ g_m(x) = c_m \\ \nabla f(x) = \lambda_1 \nabla g_1(x) + \dots + \lambda_m \nabla g_m(x). \end{cases}$$

## 8 Integration

### 8.1 Definition of Integration

**8.1.1 Geometric Motivation.** To compute the area of region under the curve  $y = f(x)$ .

- Form the upper and lower approximation:

$$U(f, \mathcal{P}) = \sum_{i=1}^n \sup_{I_i} f(x) \ell(I_i)$$

$$L(f, \mathcal{P}) = \sum_{i=1}^n \inf_{I_i} f(x) \ell(I_i).$$

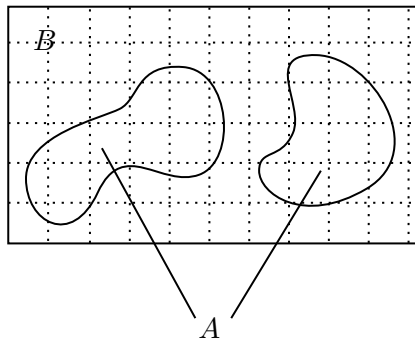
- Form the upper and lower integral:

$$\int_A^{\bar{}} f = \inf_{\mathcal{P}} U(f, \mathcal{P})$$

$$\int_A f = \sup_{\mathcal{P}} L(f, \mathcal{P}).$$

### 8.1.2 General Formulation of Integral.

- Set-up: Let  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  be a bounded function on a bounded set  $A$ .
- Goal: define the volume of the region under the surface  $y = f(x)$  (or the integral  $\int_A f \, dx$ ).
- Step 1: choose a rectangle  $B = [a_1, b_1] \times [a_2, b_2] \times \cdots \times [a_n, b_n]$  that contains  $A$ . Extend  $f$  s.t.  $f(x) = 0$  when  $x \notin A$ .



Then, the volume over  $A$  is the same as the volume over  $B$ . That is,

$$\int_A f(x) \, dx = \int_B f(x) \, dx.$$

- Step 2: partition  $B$ : divide slides of  $B$  into sub-intervals to obtain a partition  $P$ , collection of smaller rectangles.

- Step 3: Form upper and lower sums:

$$U(f, \mathcal{P}) = \sum_{R \in \mathcal{P}} \underbrace{\sup_R f(x)}_{\text{height}} \cdot \underbrace{v(R)}_{\text{base}} \quad (\text{Upper Sum of } f \text{ w.r.t. } \mathcal{P})$$

$$L(f, \mathcal{P}) = \sum_{R \in \mathcal{P}} \inf_R f(x) \cdot v(R) \quad (\text{Lower Sum of } f \text{ w.r.t. } \mathcal{P})$$

- Step 4: Form upper and lower integrals:

$$\int_A^{\bar{}} f = \inf_{\mathcal{P}} (U(f, \mathcal{P})) \quad \text{and} \quad \int_A^{\underline{}} f = \sup_{\mathcal{P}} (L(f, \mathcal{P})).$$

- Observation:

$$L(f, \mathcal{P}) \leq \text{real volume} \leq U(f, \mathcal{P}) \implies \int_A^{\underline{}} f \leq \text{real volume} \leq \int_A^{\bar{}} f.$$

- **Definition 8.1.3 (Integrable).** We say  $f$  is *Riemann integrable* if

$$\int_A^{\underline{}} f = \int_A^{\bar{}} f.$$

The integral of  $f$  on the set  $A$  is defined as  $\int_A f(x) dx = \int_A^{\underline{}} f = \int_A^{\bar{}} f$ . Sometimes, the integral is also written as  $\int_A f$  or  $\int_A f(x) dx_1 dx_2 \cdots dx_n$ .

#### Theorem 8.1.4 Equivalent Conditions for Integrability

Suppose  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  is bounded and  $A$  and  $B$  are bounded. Let  $B$  be a rectangle in  $\mathbb{R}^n$ . Assume  $f(x) = 0$  for  $x \notin A$ . Then, the following are equivalent conditions for  $f$  to be integrable:

- (Riemann's Condition):  $\forall \varepsilon > 0, \exists$  partition  $\mathcal{P}_\varepsilon$  (of  $B$ ) s.t.

$$0 \leq U(f, \mathcal{P}_\varepsilon) - L(f, \mathcal{P}_\varepsilon) < \varepsilon.$$

- (Darboux's Condition):  $\exists$  a number  $I$  s.t.  $\forall \varepsilon > 0, \exists \delta > 0$  s.t.

1.  $\mathcal{P}$  is any partition of  $B$  into rectangles  $B_1, B_2, \dots, B_N$  with side length less than  $\delta$ , and
2. If  $x_1 \in B_1, x_2 \in B_2, \dots, x_N \in B_N$ , then we have

$$\left| \sum_{i=1}^N f(x_i) v(B_i) - I \right| < \varepsilon.$$

**Remark 8.1** • The number  $I$  is the value of the integral



- $\sum_{i=1}^N f(x_i)v(B_i)$  is called the Riemann sum of  $f$  w.r.t.  $\mathcal{P}$ .
- Interpretation: Darboux's condition says that when the partition is fine enough (side length  $< \delta$ ), then the Riemann sum is a good approximation of the integral.

► **Proof 1 of Equivalent Conditions for Integrability**

**Step 1**  $f$  integrable  $\implies$  Riemann's Condition

Given  $\varepsilon > 0$ , need to find a partition  $\mathcal{P}_\varepsilon$  s.t.  $U(f, \mathcal{P}_\varepsilon) - L(f, \mathcal{P}_\varepsilon) < \varepsilon$ .

Since

$$\int_A f = \inf_{\mathcal{P}} U(f, \mathcal{P}),$$

by definition of infimum,

$$\exists \mathcal{P}_1 \text{ s.t. } U(f, \mathcal{P}_1) < \int_A f + \frac{\varepsilon}{2}.$$

Similarly,

$$\exists \mathcal{P}_2 \text{ s.t. } L(f, \mathcal{P}_2) > \int_A f - \frac{\varepsilon}{2}.$$

Let  $\mathcal{P}_\varepsilon = \mathcal{P}_1 \cup \mathcal{P}_2$  (partition refinement). Then,  $\mathcal{P}_\varepsilon$  is a refinement of  $\mathcal{P}_1$  and  $\mathcal{P}_2$ . Therefore,

$$U(f, \mathcal{P}_\varepsilon) \leq U(f, \mathcal{P}_1) < \int_A f + \frac{\varepsilon}{2}, \quad \text{and} \quad L(f, \mathcal{P}_\varepsilon) \geq L(f, \mathcal{P}_2) > \int_A f - \frac{\varepsilon}{2}.$$

Hence,

$$\begin{aligned} U(f, \mathcal{P}_\varepsilon) - L(f, \mathcal{P}_\varepsilon) &\leq U(f, \mathcal{P}_1) - L(f, \mathcal{P}_2) \\ &< \int_A f + \frac{\varepsilon}{2} - \int_A f + \frac{\varepsilon}{2} \\ &= \int_A f - \int_A f + \varepsilon \\ &= 0 + \varepsilon \\ &= \varepsilon. \quad \square \end{aligned} \quad [f \text{ integrable}]$$

**Step 2** Riemann's Condition  $\implies f$  integrable

By Assumption,  $\forall \varepsilon > 0, \exists$  partition  $\mathcal{P}_\varepsilon$  s.t.

$$U(f, \mathcal{P}_\varepsilon) - L(f, \mathcal{P}_\varepsilon) < \varepsilon.$$

Since  $\int_A f = \inf_{\mathcal{P}} U(f, \mathcal{P})$ , we have

$$\int_A f \leq U(f, \mathcal{P}_\varepsilon).$$

Similarly, we have  $\int_A f \geq L(f, \mathcal{P}_\varepsilon)$ . Then,

$$0 \leq \int_A f - \int_A f \leq U(f, \mathcal{P}_\varepsilon) - L(f, \mathcal{P}_\varepsilon) < \varepsilon.$$

Thus,

$$\int_A f = \int_A f \implies f \text{ is integrable. } \square$$

**Step 3 Darboux's Condition  $\implies$  Integrability**

Let  $I$  be the number in Darboux's condition.

$$\text{WTS: } \int_A f = I = \int_A f.$$

**Claim 8.1.5**  $\forall \varepsilon > 0, \exists$  partition  $\mathcal{P}$  s.t.

$$|L(f, \mathcal{P}) - I| < \varepsilon \quad (\star)$$

**Scratch:**

$$\begin{aligned} |L(f, \mathcal{P}) - I| &< \underbrace{\left| L(f, \mathcal{P}) - \sum_{i=1}^N f(x_i)v(B_i) \right|}_{=\sum_{i=1}^N \left| \inf_{B_i} f(x_i) - f(x_i) \right| v(B_i)} + \underbrace{\left| \sum_{i=1}^N f(x_i)v(B_i) - I \right|}_{< \frac{\varepsilon}{2}, \text{ by Darboux}} \\ &< \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon. \end{aligned}$$

So, we will make

$$\left| \inf_{B_i} f(x_i) - f(x_i) \right| < \frac{\varepsilon}{2v(B_i)N}$$

since we want  $\frac{\varepsilon}{2}$  eventually. Then,

$$|L(f, \mathcal{P}) - I| < \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon.$$

Given  $\varepsilon > 0$ . By Darboux's condition,  $\exists \delta > 0$  s.t.  $\forall \mathcal{P} = \{B_1, B_2, \dots, B_N\}$  with sides  $< \delta$ , we have

$$\left| \sum_{i=1}^N f(x_i)v(B_i) - I \right| < \frac{\varepsilon}{2}.$$

for any  $x_i \in B_i$ , where  $i = 1, \dots, N$ .

To prove  $(\star)$ , we can choose  $x_i \in B_i$  s.t.

$$0 \leq f(x_i) - \inf_{B_i} f(x_i) < \frac{\varepsilon}{2v(B_i)N}.$$

Then, it follows that

$$\begin{aligned}
 |L(f, \mathcal{P}) - I| &< \left| L(f, \mathcal{P}) - \sum_i f(x_i)v(B_i) \right| + \left| \sum_i f(x_i)v(x_i) - I \right| \\
 &< \sum_{i=1}^N \left| \inf_{B_i} f(x_i) - f(x_i) \right| v(B_i) + \frac{\varepsilon}{2} \\
 &< \sum_{i=1}^N \frac{\varepsilon}{2N \cdot \cancel{v(B_i)}} \cdot \cancel{v(B_i)} + \frac{\varepsilon}{2} \\
 &= \cancel{N} \cdot \frac{\varepsilon}{2\cancel{N}} + \frac{\varepsilon}{2} \\
 &= \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon \implies (*)
 \end{aligned}$$

Furthermore,  $(*) \implies L(f, \mathcal{P}) > I - \varepsilon \quad \forall \varepsilon > 0$ . So,

$$\int_A f = \sup_{\mathcal{P}} L(f, \mathcal{P}) \geq I.$$

Similarly,  $\forall \varepsilon > 0, \exists \mathcal{P}$  s.t.  $|U(f, \mathcal{P}) - I| < \varepsilon \implies U(f, \mathcal{P}) < I + \varepsilon$ . Then,

$$\bar{\int}_A f = \inf_{\mathcal{P}} U(f, \mathcal{P}) \leq I.$$

So, it must be

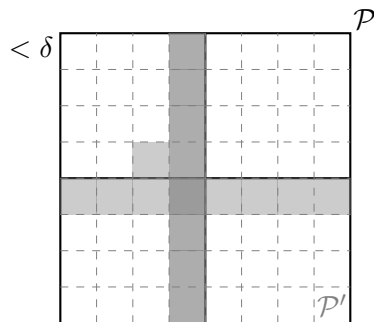
$$\bar{\int}_A f = \int_A f = I.$$


**Step 4** Integrability  $\implies$  Darboux's Condition (Scratch)

- Given  $\varepsilon > 0, \exists \mathcal{P}$  s.t.

$$I - \frac{\varepsilon}{2} < L(f, \mathcal{P}) \leq \sum_i f(x_i)v(B_i) \leq U(f, \mathcal{P}) < I + \frac{\varepsilon}{2}.$$

- Given partition  $\mathcal{P}, \exists \delta > 0$  s.t. for any partition  $\mathcal{P}'$  with side length  $< \delta$ , the sum of volumes of sub-rectangles in  $\mathcal{P}'$  that are not completely/entirely contained in a sub-rectangle in  $\mathcal{P}$  is less than  $\varepsilon$ .



 Coarse rectangle in  $\mathcal{P}$

 Fine rectangle in  $\mathcal{P}'$

 Not fully contained in  $\mathcal{P}$ , total volume  $< \varepsilon$

Q.E.D. ■

**Example 8.1.6 An Exercise**

Compute the upper and lower sums for  $\int_0^1 x \, dx$  over special partition  $\mathcal{P}$ :

$$\mathcal{P} = \left\{ 0, \frac{1}{n}, \frac{2}{n}, \dots, \frac{n-1}{n} \right\}.$$

**8.2 Criterion for Integrability**

**Question:** When is  $f$  integrable? How can we tell from other properties?

**Short Answer:**  $f$  is integrable when the set of discontinuity is “small.”

**8.2.1 Measure Zero: How to Measure the Size of a Set**

**Definition 8.2.1 (Volume of  $A$ ).** Given a bounded set  $A \subset \mathbb{R}^n$ , define *characteristic function* of  $A$  by

$$\mathbb{1}_A(x) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A \end{cases}.$$

We say that  $A$  has *volume* (or *Jordan measurable*) if  $\mathbb{1}_A(x)$  is integrable on  $A$ . We write

$$v(A) = \int_A \mathbb{1}_A(x) \, dx.$$

**Remark 8.2** When  $n = 1$ ,  $v(A)$  is the length of  $A$ . When  $n = 2$ ,  $v(A)$  is the area of  $A$ .

**Fact:** A set has volume 0 (i.e.,  $v(A) = 0$ )  $\iff \forall \varepsilon > 0, \exists$  finite cover of  $A$  by rectangles  $S_1, S_2, \dots, S_N$  s.t.

$$\sum_{i=1}^N v(S_i) < \varepsilon.$$

**Proof 1.** Suppose  $v(A) = \int_A \mathbb{1}_A(x) \, dx = 0$ . Then,  $\forall \varepsilon > 0, \exists$  partition  $\mathcal{P} = \{\mathcal{P}_1, \dots, \mathcal{P}_N\}$  of  $B$  s.t.

$$U(\mathbb{1}_A(x), \mathcal{P}) < I + \varepsilon = \varepsilon.$$

$$\implies \sum_{\mathcal{P}_j \cap A \neq \emptyset} \underbrace{\sup_{\mathcal{P}} \mathbb{1}_A(x)}_{=1} \cdot v(\mathcal{P}_j) = \sum_{\mathcal{P}_j \cap A \neq \emptyset} v(\mathcal{P}_j) < \varepsilon.$$

Note that  $\{\mathcal{P}_j \mid \mathcal{P}_j \cap A \neq \emptyset\}$  is a finite cover of  $A$ .

Q.E.D. ■

**Definition 8.2.2 (Measure Zero Set).** A set  $A \subset \mathbb{R}^n$  (not necessarily bounded) is said to have measure zero,  $m(A) = 0$ , if  $\forall \varepsilon > 0, \exists$  countable cover of  $A$  by rectangles  $\{S_i\}$  s.t.

$$\sum_{i=1}^{\infty} v(S_i) < \varepsilon.$$

**Remark 8.3**

- $v(A) = 0 \implies m(A) = 0$
- Any finite set has volume zero.
- Any countable set has measure zero. (use geometric sum: first point covered by  $\frac{\varepsilon}{2}$ , second point covered by  $\frac{\varepsilon}{4}$ , ...,  $N$ -th point covered by  $\frac{\varepsilon}{2^N}$ )

**Example 8.2.3**

Let  $A$  be the  $x$ -axis (real line).

- If  $A$  is considered as a subset of  $\mathbb{R}^2$ , then  $m(A) = 0$ .

**Proof 2.** To cover the  $x$ -axis, we can cover it interval by interval. But the volumes of the rectangles need to get smaller and smaller:

$$S_n = [n, n+1] \times \left[ -\frac{\varepsilon}{2^{|n|+2}}, \frac{\varepsilon}{2^{|n|+2}} \right]$$

for  $n = 0, \pm 1, \pm 2, \dots$

Q.E.D. ■

- However, if  $A$  is considered as a subset of  $\mathbb{R}^1$ , then  $m(A) \neq 0$ .

**Theorem 8.2.4**

Suppose  $A_i \subset \mathbb{R}^n$  with  $m(A_i) = 0 \quad \forall i = 1, 2, \dots$ . Then,

$$A = \bigcup_{i=1}^{\infty} A_i$$

has measure zero.

**Proof 3.** Given  $\varepsilon > 0$  for each  $i = 1, 2, \dots, m(A_i) = 0$ . So,  $\exists$  rectangles  $\{S_j^{(i)}\}_{j=1}^{\infty}$  s.t.  $A_i \subset \bigcup_{j=1}^{\infty} S_j^{(i)}$

with  $\sum_{j=1}^{\infty} v(S_j^{(i)}) < \frac{\varepsilon}{2^i}$ . Then,  $\{S_j^{(i)}\}_{i,j=1}^{\infty}$  is a countable collection of rectangles with

- $A = \bigcup_{i=1}^{\infty} A_i \subset \bigcup_{i=1}^{\infty} \bigcup_{j=1}^{\infty} S_j^{(i)}$

$$\bullet \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} v(S_j^{(i)}) < \sum_{i=1}^{\infty} \frac{\varepsilon}{2^i} = \varepsilon.$$

So, by definition,  $m(A) = 0$ .

Q.E.D. ■

#### Remark 8.4

- The above theorem is not true for volume zero sets. A counterexample is the rationals in  $[0, 1]$ . Each rational is volume zero, but their union is not volume zero as  $1_A$  is not integrable.
- In Definition 8.2.2, we can replace “closed rectangles  $S_i$ ” by “open rectangles  $S_i$ .”

### 8.2.2 Lebesgue's Theorem

#### Theorem 8.2.5 Lebesgue's Theorem

Let  $A$  be a bounded set in  $\mathbb{R}^n$  and  $f$  be a bounded function on  $A$ . Extend  $f$  to  $\mathbb{R}^n$  by letting  $f(x) = 0 \quad \forall x \notin A$ . Then,  $f$  is integrable on  $A \iff$  the points on which the *extended function*  $f$  is discontinuous form a set of measure zero. That is, extended  $f$  has support on  $A$ , and if  $D$  denotes the set of discontinuity of extended  $f$ , then  $m(D) = 0$ .

#### Example 8.2.6

- $A = [0, 1]$  and

$$f(x) = \begin{cases} 1 & x \text{ rational} \\ 0 & \text{o/w.} \end{cases}$$

Then, the set of discontinuity is  $D = [0, 1]$ , and  $m(D) \neq 0$ . By Lebesgue's Theorem,  $f$  is not integrable.

- $A = \{\text{rationals} \in [0, 1]\}$  and  $f(x) : A \rightarrow \mathbb{R}$  by  $f(x) \equiv 1$ . Then,  $f$  is continuous on  $A$ , but it is not integrable on  $A$ . The extended  $f$  has  $D = [0, 1]$ , not measure zero. So, by Lebesgue's Theorem,  $f$  is not integrable.
- $A = \{(x, y) \mid x^2 + y^2 < 1\} \subset \mathbb{R}^2$  and  $f(x) : A \rightarrow \mathbb{R}$  by

$$f(x, y) = \begin{cases} x^2 + \sin\left(\frac{1}{y}\right) & y \neq 0 \\ x^2 & y = 0. \end{cases}$$

Then, the set of discontinuity is  $D = [-1, 0] \times [1, 0] + \partial A$ . Then,  $m(D) = 0$  in  $\mathbb{R}^2$ . So, by Lebesgue's Theorem,  $f$  is integrable on  $A$ .

#### Corollary 8.2.7 of Lebesgue's Theorem:

- A bounded set  $A \subset \mathbb{R}^n$  has volume  $\iff \partial A$  has measure 0.

**Proof 4.** Assume  $v(A)$  exists. Then,  $\mathbb{1}_A(x)$  is integrable. So, the set of discontinuity of extended  $\mathbb{1}_A(x)$  is  $D = \partial A$ . By Lebesgue's Theorem,  $f = \mathbb{1}_A(x)$  is integrable  $\iff m(\partial A) = 0$ .

Q.E.D. ■

- Let  $A \subset \mathbb{R}^n$  be a bounded set with volume. If  $f : A \rightarrow \mathbb{R}$  is bounded and has only a (finite or) countable number of discontinuity, then  $f$  is integrable.

**Proof 5.** Denote the set of discontinuity of  $f$  on  $A$  as  $M$ . The set of discontinuity of the extended  $f$  will be  $D \subset \partial A \cup M$ . Since  $A$  has volume, by the previous Corollary, we know  $m(\partial A) = 0$ . Since  $M$  is countable,  $m(M) = 0$ . Then,  $m(\partial A \cup M) = 0 \implies D \subset \partial A \cup M$  has measure zero. By Lebesgue's Theorem,  $f$  is integrable.

Q.E.D. ■

### ► Proof 6 of Lebesgue's Theorem

#### Step 1 Preparation and Reduction

- The set-up: Fix a rectangle  $B \supset A$  (so  $\text{cl}(A) \subset \text{int}(B)$ ) and define  $g : B \rightarrow \mathbb{R}$  by

$$g(x) = \begin{cases} f(x), & x \in A \\ 0, & x \notin A. \end{cases}$$

Let  $D$  denote the set of discontinuity of  $g(x)$ . That is,

$$D = \{x \in B \mid g(x) \text{ is not continuous at } x\}.$$

Need to show:  $f$  integrable on  $A \iff m(D) = 0$ .

- How to quantify discontinuity?

1. **Definition 8.2.8 (Oscillation).** The *oscillation* of a function  $h(x)$  at a point  $x_0$  is

$$\mathcal{O}(h, x_0) = \inf \left\{ \sup \{ |h(x_2) - h(x_1)| : x_1, x_2 \in U \} : U \text{ is a neighborhood of } x_0 \right\},$$

where  $\mathcal{O}(f, U) = \sup \{ |h(x_2) - h(x_1)| : x_1, x_2 \in U \}$  is the oscillation in a neighborhood  $U$ , and  $\inf$  takes over all possible neighborhoods of  $x_0$ .

2. **Claim 8.2.9**  $h$  is continuous at  $x_0 \implies \mathcal{O}(h, x_0) = 0$ .

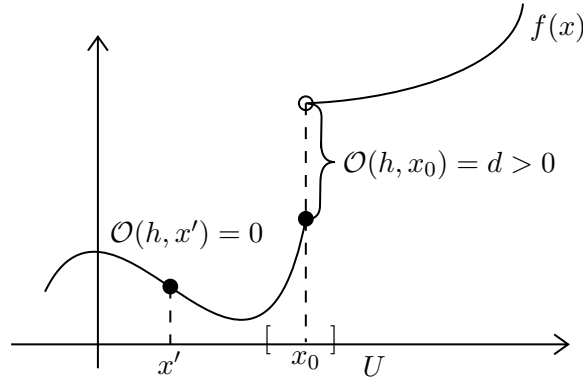
*Proof.*  $h$  is continuous at  $x_0 \implies \forall \varepsilon > 0, \exists \delta > 0$  s.t.

$$|x - x_0| < \delta \implies |h(x) - h(x_0)| < \frac{\varepsilon}{2}.$$

For  $U = \{|x - x_0| < \delta\} \cap A$ ,

$$\begin{aligned} x_1, x_2 \in U &\implies |h(x_2) - h(x_1)| \leq |h(x_2) - h(x_0)| + |h(x_0) - h(x_1)| \\ &< \frac{\varepsilon}{2} + \frac{\varepsilon}{2} = \varepsilon. \end{aligned}$$

Then,  $\mathcal{O}(h, U) < \varepsilon \implies \mathcal{O}(h, x_0) = 0$ .  $\square$



**Step 2** ( $\Leftarrow$ ) **Assume  $m(D) = 0$ . Prove  $g$  is integrable.**

We will show:  $g$  satisfies Riemann condition.

• Set up:

Fix  $\varepsilon > 0$ . Let  $D_\varepsilon = \{x \in B \mid \mathcal{O}(g, x) > \varepsilon\}$ . Then,  $D_\varepsilon \subset D$ . So,  $m(D_\varepsilon) = 0$ .

By Definition,  $\exists$  collection of open rectangles  $|B_i|$  s.t.

$$D_\varepsilon \subset \bigcup_i B_i \quad \text{and} \quad \sum_i v(B_i) < \varepsilon.$$

**Claim 8.2.10**  $D_\varepsilon$  is closed (and hence compact).

*Proof.* (Sketch)  $D_\varepsilon$  contains all its limits points. That is,

$$x_n \in D_\varepsilon, \{x_n\} \rightarrow x \implies x \in D_\varepsilon.$$

Assume, for the sake of contradiction,

$$x \notin D_\varepsilon \implies \mathcal{O}(g, x) < \varepsilon.$$

But  $\mathcal{O}(g, x_n) \geq \varepsilon$ , we can derive a contradiction from there.  $\square$

Since  $D_\varepsilon$  is compact, it has a finite subcover:

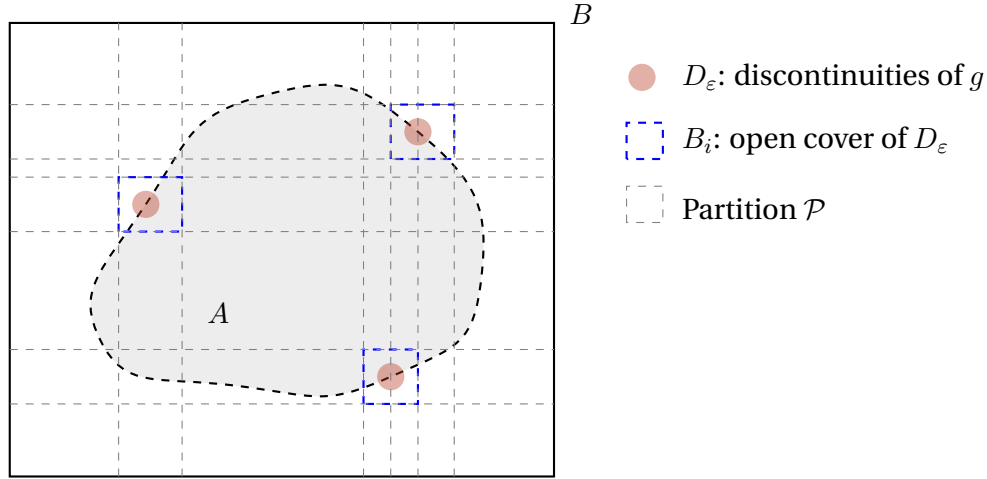
$$\{B_1, B_2, \dots, B_N\} \text{ s.t. } \sum_{i=1}^N v(B_i) < \varepsilon.$$



- Initial Partition of  $B$ :

Construct a partition  $\mathcal{P}$  from  $\{B_i\}_{i=1}^N$  s.t. each rectangle  $S \in \mathcal{P}$  is either:

1. disjoint from  $D_\varepsilon$ , or
2. its interior is contained in one of the  $B_i$ 's.



The way to construct  $\mathcal{P}$  is to extend the sides of  $B_i$  to form a partition on  $B$ .

Let  $C_1 = \{S \in \mathcal{P} : \text{int}(S) \text{ is contained in one of } B_i\}$  and  $C_2 = \{S \in \mathcal{P} : S \cap D_\varepsilon = \emptyset\}$ .

- Refinement of  $\mathcal{P}$

Fix  $S \in C_2$ ,  $S \cap D_\varepsilon = \emptyset \implies \mathcal{O}(g, x) < \varepsilon \quad \forall x \in S$ . Then,  $\forall x \in S, \exists$  neighborhood  $U_x$  s.t.

$$\sup \{|g(x_1) - g(x_2)| : x_1, x_2 \in U_x\} < \mathcal{O}(g, x) + \delta,$$

where  $\delta = \frac{1}{2}(\varepsilon - \mathcal{O}(g, x))$ . Then,

$$\sup_{U_x} g - \inf_{U_x} g < \mathcal{O}(g, x) + 2\delta = \varepsilon.$$

Denote  $M_{U_x}(g) = \sup_{U_x} g$  and  $m_{U_x}(g) = \inf_{U_x} g$ . Then,

$$\boxed{M_{U_x}(g) - m_{U_x}(g) < \varepsilon} \quad (\star)$$

Since  $S$  is compact and  $S \subset \bigcup_{x \in S} U_x$ .

$\implies \exists$  finite collection of neighborhoods  $\{U_{x_i}\}$  that covers  $S$ . Partition  $S$  so that each rectangle is contained in some  $U_{x_i}$ . Do this partition for each  $S \in C_2$ , and we obtain a refinement of  $\mathcal{P}$ , denoted by  $\mathcal{P}'$ .

- Verify Riemann's condition for  $\mathcal{P}'$ :

Note that

$$\begin{aligned}
 U(g, \mathcal{P}') - L(g, \mathcal{P}') &= \sum_{S' \in \mathcal{P}'} (M_{S'}(g) - m_{S'}(g))v(S') \\
 &= \sum_{S' \subset S \in C_1} (M_{S'}(g) - m_{S'}(g))v(S') + \sum_{S' \subset S \in C_2} (M_{S'}(g) - m_{S'}(g))v(S') \\
 &\leq \sum_{S' \subset S \in C_1} 2Mv(S') + \sum_{S' \subset S \in C_2} \varepsilon v(S') \quad [|g(x)| \leq M \text{ and } (\star)] \\
 &\leq 2M \sum_i v(B_i) + \varepsilon v(B) \quad [C_1 \text{ is covered by } B'_i s] < 2M\varepsilon + \varepsilon v(B) \quad [L] \\
 &= \varepsilon(2M + v(B)).
 \end{aligned}$$

In summary, given  $\varepsilon > 0$ ,  $\exists$  partition  $\mathcal{P}'$  s.t.

$$U(g, \mathcal{P}') - L(g, \mathcal{P}') < \varepsilon(2M + v(B)).$$

So, we satisfy Riemann condition.  $\square$

**Step 3**  $(\Rightarrow) f \text{ is integrable} \Rightarrow m(D) = 0$ .

For  $n = 1, 2, \dots$ , let

$$D_{1/n} = \left\{ x \in D \mid \mathcal{O}(g, x) \geq \frac{1}{n} \right\}.$$

Then,

$$D = \bigcup_{i=1}^{\infty} D_{1/n}.$$

**Need to show:**  $m(D_{1/n}) = 0 \quad \forall n$ .

Fix  $n \geq 1$ . For any partition  $\mathcal{P}$ , write

$$D_{1/n} = S_1 \cup S_2,$$

where

$$S_1 = \{x \in D_{1/n} \mid x \text{ is on the boundary of some } S \in \mathcal{P}\}$$

and

$$S_2 = \{x \in D_{1/n} \mid x \in \text{int}(S) \text{ for some } S \in \mathcal{P}\}.$$

Then,  $m(S_1) = 0$ . We need to show  $m(S_2) = 0$ .

Given  $\varepsilon > 0$ . By Riemann's condition,  $\exists$  partition  $\mathcal{P}$  s.t.

$$\sum_{S \in \mathcal{P}} (M_S(g) - m_S(g))v(S) < \frac{\varepsilon}{n}.$$

Let  $C$  denote the collection of all  $S \in \mathcal{P}$  s.t.  $D_{1/n} \cap \text{int}(S) \neq \emptyset$ . Then,  $C$  covers  $S_2$  and for any  $S \in C$ ,

$$M_S(g) - m_S(g) \geq \mathcal{O}(g, x) \geq \frac{1}{n}.$$

Thus,

$$\sum_{S \in C} (M_S(g) - m_S(g))v(S) \leq \sum_{S \in \mathcal{P}} (M_S(g) - m_S(g))v(S) < \frac{\varepsilon}{n}.$$

Since

$$\sum_{S \in C} (M_S(g) - m_S(g))v(S) \geq \sum_{S \in C} \frac{1}{n}v(S) = \frac{1}{n} \sum_{S \in C} v(S),$$

we have

$$\frac{1}{n} \sum_{S \in C} v(S) \leq \sum_{S \in C} (M_S(g) - m_S(g))v(S) < \frac{\varepsilon}{n}.$$

That is,

$$\frac{1}{n} \sum_{S \in C} v(S) < \frac{\varepsilon}{n} \implies \sum_{S \in C} v(S) < \varepsilon.$$

Therefore,  $m(S_2) = 0$  as well.

Since  $m(S_1) = m(S_2) = 0$  and  $D_{1/n} = S_1 \cup S_2$ ,  $m(D_{1/n}) = 0 \quad \forall n$ . Then,

$$m(D) = m\left(\bigcup_{i=1}^{\infty} D_{1/n}\right) = 0.$$

Q.E.D. ■

### Theorem 8.2.11 Properties of Integration

Let  $A, B \subset \mathbb{R}^n$  be bounded,  $c \in \mathbb{R}$ , and  $f, g : A \rightarrow \mathbb{R}$  be integrable. Then,

- $f + g$  is integrable and  $\int_a (f + g) = \int_A f + \int_A g$ .
- $cf$  is integrable and  $\int_A (cf) = c \int_A f$ .
- $|f|$  is integrable and  $\left| \int_A f \right| \leq \int_A |f|$ .
- If  $f \leq g$ , then  $\int_A f \leq \int_A g$ .
- If  $A$  has volume and  $|f| \leq M$ , then  $\left| \int_A f \right| \leq Mv(A)$ .
- **(Mean Value Theorem for Integrals):** If  $f : A \rightarrow \mathbb{R}$  is continuous and  $A$  has volume and is compact and connected, then  $\exists x_0 \in A$  s.t.  $\int_A f(x) dx = f(x_0)v(A)$ . The quantitive  $\frac{1}{v(A)} \cdot \int_A f$  is called the *average* of  $f$  over  $A$ .
- Let  $f : A \cup B \rightarrow \mathbb{R}$ . If the sets  $A$  and  $B$  are such that  $A \cap B$  has measure zero and  $f|_{(A \cap B)}$ ,  $f|_A$ , and  $f|_B$  are all integrable, then  $f$  is integrable on  $A \cup B$  and  $\int_{A \cup B} = \int_A f + \int_B f$ .

### 8.3 Improper Integrals

**Goal:** Study integral of the form  $\int_A f(x)$ , where  $f : A \subset \mathbb{R}^n \rightarrow \mathbb{R}$  is an arbitrary function and  $A \subset \mathbb{R}^n$  is an arbitrary set.

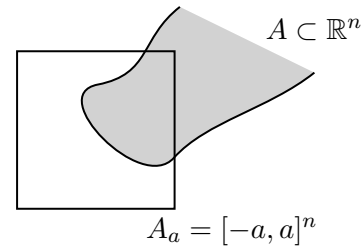
**Definition 8.3.1 (Integral).**

- If  $A \subset \mathbb{R}^n$  is bounded and  $f$  is bounded, then

$$\int_A f(x) = \bar{\int}_A f(x) = \underline{\int}_A f(x) \quad (\text{Riemann Condition})$$

- $f(x) \geq 0$  bounded and  $A$  is arbitrary, then

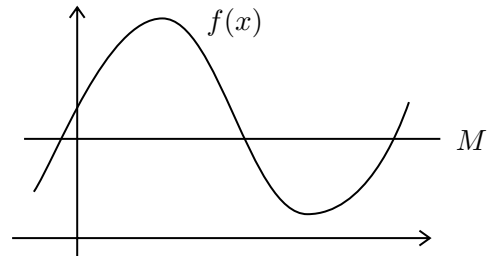
$$\int_A f(x) = \lim_{a \rightarrow \infty} \int_{A_a} f(x)$$



- $f(x) \geq 0$  unbounded and  $A$  is arbitrary.

For  $M > 0$ , define

$$f_M(x) = \begin{cases} f(x) & \text{for } f(x) \leq M \\ 0 & \text{o/w.} \end{cases}$$



Then,

$$\int_A f(x) = \lim_{M \rightarrow \infty} \int_A f_M(x).$$

- $f$  is arbitrary and  $A$  is arbitrary.

Let

$$f^+(x) = \begin{cases} f(x) & f(x) \geq 0 \\ 0 & f(x) < 0, \end{cases} \quad \text{and} \quad f^-(x) = \begin{cases} 0 & f(x) \geq 0 \\ -f(x) & f(x) < 0. \end{cases}$$

**Remark 8.5** 1.  $f^+(x)$  is the positive part of  $f$  and  $f^-(x)$  is the negative part of  $f$ .

2.  $f^+, f^- \geq 0$ .

3.  $f(x) = f^+(x) - f^-(x)$ . We can write any function as the difference of two non-negative functions.

$$4. |f(x)| = f^+(x) + f^-(x).$$

So,  $f$  is integrable on  $A$  if both  $f^+$  and  $f^-$  are integrable on  $A$ . We write

$$\int_A f(x) = \int_A f^+(x) - \int_A f^-(x).$$

**Remark 8.6** 1. One can show this definition preserves linearity of integral from bounded case.

2. **Observation:**  $f$  integrable  $\implies f^+$  and  $f^-$  integrable  $\implies |f| = f^+ + f^-$  is also integrable.

However,  $|f|$  integrable  $\not\Rightarrow f$  integrable. For counterexample,

$$f(x) = \begin{cases} 1 & x \text{ rational} \\ -1 & x \text{ irrational} \end{cases} \quad \text{on } [0, 1].$$

$|f(x)| \equiv 1 \implies$  integrable. But  $f^+$ ,  $f^-$ , or  $f$  are not integrable.

### Theorem 8.3.2 Comparison Principle

Suppose

- $0 \leq g \leq f$  on  $A$  and  $\int_A f$  converges (i.e.,  $f$  is integrable on  $A$ )
- $g$  is integrable on each finite rectangle  $[-a, a]^n$ .

Then,  $g$  is also integrable on  $A$ , and  $\int_A g \leq \int_A f$ .

**Remark 8.7** The second condition is crucial and cannot be removed.

**Proof 1.** Since  $g \geq 0$  and is integrable on  $[-a, a]^n$ , define

$$G(a) = \int_{[-a, a]^n} g(x).$$

Then,  $G(a)$  is an increasing function of  $a$ . Furthermore,

$$g \leq f \implies G(a) = \int_{[-a, a]^n} g(x) \leq \int_{[-a, a]^n} f(x) \leq \int_A f(x).$$

So,

$$\int_A g(x) = \lim_{a \rightarrow \infty} G(a) \leq \int_A f(x).$$

Q.E.D. ■

**Question:** When does an integrable  $\int_a^b f(x)$  (one-variable function) converge? If it converges, how to compute?

**Theorem 8.3.3 Integral of Functions of One-Variable**

- Suppose  $f : [a, \infty) \rightarrow \mathbb{R}$  is continuous with  $f(x) \geq 0$ . Let  $F'(x) = f(x)$ . Then,

$$\int_a^\infty f(x) \, dx = \lim_{b \rightarrow \infty} \int_a^b f(x) \, dx = \lim_{b \rightarrow \infty} [F(b) - F(a)].$$

- Suppose  $f : (a, b] \rightarrow \mathbb{R}$  is continuous with  $f(x) \geq 0$ . Then,

$$\int_a^b f(x) \, dx = \lim_{\varepsilon \rightarrow 0^+} \int_{a+\varepsilon}^b f(x) \, dx.$$

**Example 8.3.4**

- Consider  $\int_1^\infty x^p \, dx$ .

**Solution 2.**For  $b \geq 1$ ,

$$\int_1^b x^p \, dx = \begin{cases} \ln b & p = -1 \\ \frac{1}{p+1}(b^{p+1} - 1) & p \neq -1. \end{cases}$$

When  $b \rightarrow \infty$ ,  $\int_1^b x^p \, dx$  diverges when  $p \geq -1$  and converges when  $p < -1$ . So,

$$\int_1^\infty x^p \, dx \quad \text{is divergent when } p \geq -1$$

and

$$\int_1^\infty x^p \, dx = -\frac{1}{p+1} \quad \text{is convergent when } p < -1.$$

□

- Consider  $\int_1^\infty e^{-x^2} x^3 \, dx$ .

**Solution 3.**

Converges by comparison.

□

**Definition 8.3.5 (Conditional Convergence).**

$$\int_a^\infty f(x) \, dx \quad (\text{conditional}) = \lim_{b \rightarrow \infty} \int_a^b f(x) \, dx.$$

**Remark 8.8 (Types of Convergence)** For an improper integral  $\int_a^\infty f(x) \, dx$ , there are three types of convergence:

- **Absolute Convergence:**  $\int_a^\infty |f(x)| \, dx$  exists.
- **Conditional Convergence:**  $\lim_{b \rightarrow \infty} \int_a^b f(x) \, dx$  exists.
- **Divergence.**

For general function, absolute convergence  $\not\Rightarrow$  conditional convergence. For continuous function, absolute convergence is stronger, and  $\Rightarrow$  conditional convergence.

### Example 8.3.6

Determine whether the integral  $\int_1^\infty \frac{\cos x}{x} \, dx$  is absolute convergence, conditional convergence, or neither (divergence).

#### Solution 4.

- First, consider absolute convergence.

Observe that

$$\begin{aligned} \int_0^\infty \left| \frac{\cos x}{x} \right| \, dx &= \int_1^\infty \frac{|\cos x|}{x} \, dx \geq \int_{\pi/2}^{n\pi/2} \frac{|\cos x|}{x} \, dx \\ &= \sum_{k=1}^{n-1} \int_{k\pi/2}^{(k+1)\pi/2} \frac{|\cos x|}{x} \, dx \\ &\geq \sum_{k=1}^{n-1} \underbrace{\frac{1}{(k+1)\frac{\pi}{2}}}_{\text{harmonic}} \int_{k\pi/2}^{(k+1)\pi/2} |\cos x| \, dx \\ &\rightarrow \infty \quad \text{as } n \rightarrow \infty. \end{aligned}$$

So,  $\int_1^\infty \left| \frac{\cos x}{x} \right| \, dx$  diverges, and thus  $\int_1^\infty \frac{\cos x}{x} \, dx$  is not absolutely convergent.

- Conditional convergence:

$$\int_1^b \frac{\cos x}{x} \, dx = \frac{\sin x}{x} \Big|_1^b + \int_1^b \frac{\sin x}{x^2} \, dx \quad [\text{Integration by Parts}]$$

When  $b \rightarrow \infty$ ,

$$\lim_{b \rightarrow \infty} \frac{\sin x}{x} \Big|_1^b = \frac{\sin 1}{1} \quad \text{converges.}$$

Further,

$$\left| \frac{\sin x}{x^2} \right| \leq \left| \frac{1}{x^2} \right| = \frac{1}{x^2} \Rightarrow \int_1^\infty \left| \frac{\sin x}{x^2} \right| \, dx \leq \int_1^\infty \frac{1}{x^2} \, dx.$$

So,  $\int_1^\infty \frac{\sin x}{x^2} dx$  absolutely converges by comparison.

Then,  $\int_1^b \frac{\cos x}{x} dx$  is conditional convergence.

□

## 8.4 Lebesgue Convergence Theorem

**Goal:** When do we have

$$\lim_{n \rightarrow \infty} \int_A f(x) dx = \int_A \left( \lim_{n \rightarrow \infty} f(x) \right) dx? \quad (\star)$$

### Theorem 8.4.1 Lebesgue Monotone Convergence Theorem (LMCT)

Let  $g_n : [0, 1] \rightarrow \mathbb{R}$  be a sequence of non-negative integrable function such that

- $g_{n+1}(x) \leq g_n(x) \quad \forall x \in [0, 1]$  (decreasing sequence)
- $\lim_{n \rightarrow \infty} g_n(x) = 0 \quad \forall x \in [0, 1]$ .

Then,

$$\lim_{n \rightarrow \infty} \int_0^1 g_n(x) dx = \int_0^1 0 dx = 0.$$

**Corollary 8.4.2 :** Suppose  $f_n(x), f(x) : [0, 1] \rightarrow \mathbb{R}$  with

- $f_n \leq f_{n+1}(x) \leq f(x) \quad \forall x \in [0, 1]$
- $f_n(x) \rightarrow f(x) \quad \forall x$ .

Then,

$$\lim_{n \rightarrow \infty} \int_0^1 f_n(x) dx = \int_0^1 f(x) dx.$$

**Proof 1.** Apply LMCT to the sequence  $g_n(x) = f(x) - f_n(x) \geq 0$ .

Q.E.D. ■

### Remark 8.9

- For  $(\star)$  to hold, we only need  $f_n(x) \uparrow f(x)$  ( $f_n(x)$  is monotone increasing and the limit of  $f_n(x)$  is  $f(x)$ )
- The assumption that  $A = [0, 1] \subset \mathbb{R}$  is not essential. Result is true for any set  $A \subset \mathbb{R}^n$ .
- The monotonicity assumption cannot be removed. For example:

$$g_n(x) = \begin{cases} n, & 0 < x < \frac{1}{n} \\ 0, & \text{o/w} \end{cases}$$



Then, we have  $g_n(x) \rightarrow 0 \quad \forall x \in [0, 1]$ . However,

$$\int_0^1 g_n(x) dx = 1 \quad \forall n \quad \text{and} \quad \int_0^1 0 dx = 0.$$

So,

$$\int_0^1 g_n dx \neq \int_0^1 0 dx,$$

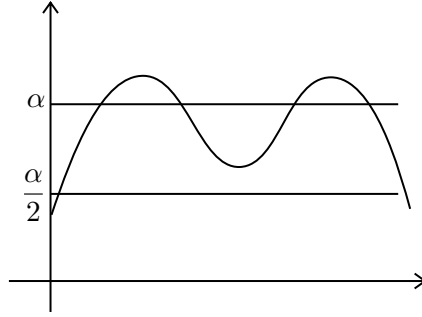
and LMCT does not hold anymore.

### ► Proof 2 of Lebesgue Monotone Convergence Theorem

**Lemma 8.4.3 :** Suppose  $f : [0, 1] \rightarrow \mathbb{R}$  be integrable with  $|f| \leq M$  and  $\int_0^1 f \geq \alpha > 0$ . Then, the set

$$E = \left\{ x \in [0, 1] \mid f(x) \geq \frac{\alpha}{2} \right\}$$

contains a finite union of disjoint open intervals of total length  $\geq \frac{\alpha}{4M}$ .



*Proof.* By definition of integral,  $\exists$  partition  $\mathcal{P}$  s.t.

$$0 \leq \int_0^1 f - L(f, \mathcal{P}) < \frac{\alpha}{4}.$$

Then,

$$L(f, \mathcal{P}) > \int_0^1 f - \frac{\alpha}{4} \geq \alpha - \frac{\alpha}{4} = \frac{3\alpha}{4}.$$

Let  $\ell$  denote the total length of the intervals  $I$  in  $\mathcal{P}$  with  $I \subset E$ . Then,

$$\begin{aligned} \frac{3\alpha}{4} &< L(f, \mathcal{P}) = \sum_{I \in \mathcal{P}} \left( \inf_I f(x) \right) \ell(I) \\ &= \sum_{I \in \mathcal{P} \cap E} \left( \inf_I f(x) \right) \ell(I) + \sum_{I \in \mathcal{P} \setminus E} \left( \inf_I f(x) \right) \ell(I) \\ &\leq \sum_{I \in \mathcal{P} \cap E} M \cdot \ell(I) + \sum_{I \in \mathcal{P} \setminus E} \frac{\alpha}{2} \ell(I) \\ &\leq \ell M + \frac{\alpha}{2} \cdot 1 \end{aligned} \quad \left[ \text{If } I \notin E, f(x) \leq \frac{\alpha}{2} \right]$$

So,  $\ell \cdot M \geq \frac{\alpha}{4} \implies \ell \geq \frac{\alpha}{4M}$ . Remove endpoints from  $I$ , we get open intervals.  $\square$

- **Step 1** Set up and Reduction:

$$0 \leq g_{n+1} \leq g_n \implies \int_0^1 g_{n+1}(x) dx \leq \int_0^1 g_n(x) dx.$$

Then, the limit exists:

$$\lim_{n \rightarrow \infty} \int_0^1 g_n(x) dx =: \lambda \geq 0.$$

Need to show:  $\lambda = 0$ .

Assume  $\lambda > 0$ , and we will derive a contradiction (with the assumption  $g_n(x) \rightarrow 0 \quad \forall x \in [0, 1]$ ).

- **Step 2** Apply the above Lemma 8.4.3 to the cut-off function  $(g_n)_M$ , where  $M > 0$ .

$$(g_n(x))_M := \begin{cases} g_n(x), & g_n(x) \leq M \\ M, & g_n(x) > M. \end{cases}$$

Then,

$$\int_0^1 g_n(x) dx = \lim_{M \rightarrow \infty} \int_0^1 (g_n)_M.$$

Choose  $M = \frac{2\lambda}{5}$  s.t.

$$0 \leq \int_0^1 (g_n - (g_n)_M) \leq \int_0^1 (g_1 - (g_1)_M) \leq \frac{\lambda}{5}.$$

Let  $E_n = \left\{ x \in [0, 1] \mid g_n(x) \geq \frac{2\lambda}{5} \right\}$ . Then,

1.  $E_{n+1} \subset E_n$  by monotonicity
2.  $\left\{ x \in [0, 1] \mid (g_n)_M(x) \geq \frac{\alpha}{2} \right\} \subset E_n$ . Choose  $\alpha$  s.t.  $\frac{2\lambda}{5} = \frac{\alpha}{2}$  to apply the Lemma.  $\implies \alpha = \frac{4\lambda}{5}$ .

Apply Lemma 8.4.3 to  $(g_n)_M$  and  $\alpha = \frac{4\lambda}{5}$ . Then,  $E_n$  contains a finite union of disjoint open intervals of total length

$$\ell \geq \frac{\alpha}{4M} = \frac{4\lambda}{5} \cdot \frac{1}{4M} = \frac{\lambda}{5M}$$

- **Step 3** Show that  $\bigcap_{n=1}^{\infty} E_n \neq \emptyset$ .

Let

$$D = \bigcup_{n=1}^{\infty} \{x \in [0, 1] \mid g_n \text{ not continuous at } x\} = \bigcup_{n=1}^{\infty} D_n.$$

Since  $g_n$  is integrable, we have  $m(D_n) = 0$ . So,

$$m(D) = m\left(\bigcup_{n=1}^{\infty} D_n\right) = 0.$$

That is,  $D$  is covered by  $U$ , a countable union of open intervals of total length  $< \varepsilon = \frac{\lambda}{5M}$ .

By Step 2,  $E_n \not\subset U$ .

**Claim 8.4.4**  $\text{cl}(E_n) \subset E_n \cup U$ .

*Proof.* In fact, if  $x_0 \in \text{cl}(E_n) \setminus E_n$ , then [WTS:  $x_0 \in U$ ]

$$g_n(x_0) < \frac{2\lambda}{5} \implies g_n \text{ is not continuous at } x_0.$$

Suppose  $x_0 \in \text{cl}(E_n) \implies \exists x_k \in E_n$  s.t.  $x_k \rightarrow x_0$  as  $k \rightarrow \infty$ . Also,  $g_n(x_k) \geq \frac{2\lambda}{5}$ , but  $g_n(x_0) < \frac{2\lambda}{5} \implies g_n(x_k) \neq g_n(x_0) \implies \text{discontinuous}$

So,  $x_0 \in D_n$ , and thus  $x_0 \in U$ . So, this Claim 8.4.4 is true.  $\square$

Note, let  $F_n = \text{cl}(E_n) \setminus U$ . Then,

1.  $F_n$  is compact
2.  $F_n \subset E_n$  (by Claim 8.4.4)

So, by the nested set property:  $\bigcap_{n=1}^{\infty} F_n \neq \emptyset$ . As  $F_n \subset E_n$ , we further have  $\bigcap_{n=1}^{\infty} E_n \neq \emptyset$ .

Let  $x_0 \in \bigcap_{n=1}^{\infty} E_n$ , then  $g_n(x_0) \geq \frac{2\lambda}{5}$ . Then,  $\lim_{n \rightarrow \infty} g_n(x_0) \neq 0$ . \*This derives a contradiction with the second assumption in LMCT (i.e.,  $g_n(x) \rightarrow 0$ ). So,  $\lambda > 0$  is impossible, and it must be that  $\lambda = 0$ .

Q.E.D.  $\blacksquare$

**Corollary 8.4.5:** Let  $g_n : A \rightarrow \mathbb{R}$  be integrable and non-negative. Assume

$$g(x) = \sum_{n=1}^{\infty} g_n(x)$$

is also integrable. Then,

$$\int_A g(x) = \int_A \sum_{n=1}^{\infty} g_n(x) = \sum_{n=1}^{\infty} \int_A g_n(x).$$

**Proof3.** Let  $f_n(x) = \sum_{k=1}^n g_k(x)$ , the partial sum.

Then,

$$\int_A f_n(x) = \int_A \sum_{k=1}^n g_k(x) = \sum_{k=1}^n \int_A g_k(x) \quad [\text{property of integral}]$$

As  $n \rightarrow \infty$ ,  $f_n \rightarrow g(x)$ , and  $f_{n+1} \geq f_n$  ( $g_n$  is non-negative). Then, apply Corollary 8.4.2, we have

$$\int_A g(x) = \sum_{n=1}^{\infty} \int_A g_n(x).$$

Q.E.D.  $\blacksquare$

## 9 Computing Integrals

**Question:** In practice, how do we compute the integral  $\int_A f(x) \, dx$ ?

- In  $\mathbb{R}^1$ : Fundamental Theorem of Calculus.

$$\int_a^b f(x) \, dx = F(x) \Big|_a^b = F(b) - F(a).$$

- In  $\mathbb{R}^n$ : Reduce to  $\mathbb{R}^1$  case by *Fubini's Theorem*. Or, use *change of variable* (substitution first), and then use Fubini's Theorem.

### 9.1 Fubini's Theorem

#### Theorem 9.1.1 Fubini's Theorem

Let  $A = \{(x, y) \mid a \leq x \leq b, c \leq y \leq d\}$  be a rectangle in  $\mathbb{R}^2$  and  $f : A \rightarrow \mathbb{R}$  be integrable. Suppose for each fixed  $x \in [a, b]$ , the following integral exists:

$$g(x) = \int_c^d f(x, y) \, dy.$$

Then,  $g(x)$  is integrable on  $[a, b]$ , and

$$\int_A f(x, y) = \int_a^b g(x) \, dx = \int_a^b \left( \int_c^d f(x, y) \, dy \right) dx.$$

**Corollary 9.1.2:** If  $f : A \rightarrow \mathbb{R}$  is continuous, then

$$\int_A f(x, y) = \int_a^b \left( \int_c^d f(x, y) \, dy \right) dx \stackrel{\text{symmetry}}{=} \int_c^d \left( \int_a^b f(x, y) \, dx \right) dy.$$

**Corollary 9.1.3 Generalization:** Let  $A$  be a region given by  $A = \{(x, y) \mid a \leq x \leq b, \varphi(x) \leq y \leq \psi(x)\}$ , where  $\varphi$  and  $\psi$  are continuous. If  $f : A \rightarrow \mathbb{R}$  is continuous, then

$$\int_A f(x, y) = \int_a^b \left( \int_{\varphi(x)}^{\psi(x)} f(x, y) \, dy \right) dx.$$

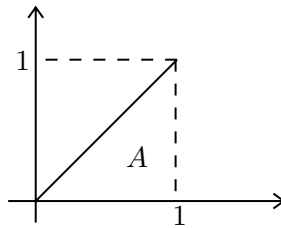
#### Remark 9.1

- The roles of  $x$  and  $y$  can be interchanged.
- Results are true in higher dimensions. For example, let  $C = A \times B \subset \mathbb{R}^{n+m}$ , where  $A \subset \mathbb{R}^n$  and  $B \subset \mathbb{R}^m$ . Fix  $x \in A$  and  $y \in B$ . Then,

$$\int_{A \times B} f = \int_A \left( \int_B f(x, y) \, dy \right) dx.$$

**Example 9.1.4 Computing Integral**

Compute  $\int_A (x + y) \, dx \, dy$ , where  $A$  is the following region:

**Solution 1.**

$$\begin{aligned}
 \int_A (x + y) \, dx \, dy &= \int_0^1 \left( \int_0^x (x + y) \, dy \right) dx \\
 &= \int_0^1 \left( xy + \frac{1}{2}y^2 \right) \Big|_0^1 dx \\
 &= \int_0^1 x^2 + \frac{1}{2}x^2 \, dx \\
 &= \frac{3}{2} \cdot \frac{1}{3} x^3 \Big|_0^1 \\
 &= \frac{1}{2}.
 \end{aligned}$$

□

**► Proof 2 of Fubini's Theorem**

- Let  $g(x) = \int_c^d f(x, y) \, dy$ . **WTS:** (1)  $g$  is integrable on  $[a, b]$ , and (2)  $\int_a^b g \, dx = \int_A f$ . We will compute the upper and lower sums of  $f$  and  $g$ .
- Fix any partition  $\mathcal{P}_A$  of  $A$ , where  $\mathcal{P}_A = \{S_{i,j}\}_{i,j}$ , where  $S_{i,j} = v_i \times w_j$ . Then,  $\mathcal{P}_A$  induces a partition of  $[a, b]$ , where  $\mathcal{P}_{[a,b]} = \{v_i\}_i$  and a partition of  $[c, d]$ ,  $\mathcal{P}_{[c,d]} = \{w_j\}_j$ .
- Next, estimate the lower sum  $L(f, \mathcal{P}_A)$ :

$$\begin{aligned}
 L(f, \mathcal{P}_A) &= \sum_{i,j} \underbrace{\inf_{x \in S_{i,j}} f(x)}_{\text{denote as } m_{i,j}(f)} v(S_{i,j}) \\
 &= \sum_{i,j} m_{i,j}(f) v(v_i \times w_j) \\
 &= \sum_{i,j} m_{i,j}(f) v(v_i) \cdot v(w_j).
 \end{aligned}$$

**Key Observation:**

$$\inf \{f(x, y) \mid (x, y) \in v_i \times w_j\} \leq \underbrace{\inf \{f(x, y) : y \in w_j\}}_{\text{fix } x, \text{ allow } y \text{ to vary}} \quad \forall x \in v_i.$$

Denote  $\inf \{f(x, y) \mid y \in w_j\} = m_j(f, x)$ . Then, for any fixed  $x \in [a, b]$ ,

$$\begin{aligned} m_{i,j}(f) &\leq m_j(f, x) \\ m_{i,j}(f)v(w_j) &\leq m_j(f, x)v(w_j) \\ \sum_j m_{i,j}(f)v(w_j) &\leq \underbrace{\sum_j m_j(f, x) \cdot v(w_j)}_{\text{lower sum of } f(x,y) \text{ in the} \\ &\quad \text{variable } y \text{ w.r.t. partition } \mathcal{P}_{[c,d]}} \\ &= L(f(x, y), \mathcal{P}_{[c,d]}) \\ &\leq \int_c^d f(x, y) dy = g(x) \quad \forall x. \end{aligned}$$

Thus,

$$\begin{aligned} \sum_j m_{i,j}(f)v(w_j) &\leq \inf_{v_i} g(x) \\ \sum_j m_{i,j}(f)v(w_j)v(v_i) &\leq \inf_{v_i} g(x)v(v_i) \\ \sum_i \sum_j m_{i,j}(f)v(w_j)v(v_i) &\leq \sum_i \inf_{v_i} g(x)v(v_i) \\ \underbrace{\sum_{i,j} m_{i,j}(f)v(w_j)v(v_i)}_{L(f, \mathcal{P}_A)} &\leq \underbrace{\sum_i \inf_{v_i} g(x)v(v_i)}_{L(g, \mathcal{P}_{[a,b]})} \end{aligned}$$

So,

$$L(f, \mathcal{P}_A) \leq L(g, \mathcal{P}_{[a,b]}).$$

- Similarly, we have

$$U(f, \mathcal{P}_A) \geq U(g, \mathcal{P}_{[a,b]}).$$

- Therefore, we have

$$L(f, \mathcal{P}_A) \leq L(g, \mathcal{P}_{[a,b]}) \leq U(g, \mathcal{P}_{[a,b]}) \leq U(f, \mathcal{P}_A).$$

Since  $f$  is integrable, by Riemann's condition,

$$0 \leq U(f, \mathcal{P}_A) - L(f, \mathcal{P}_A) < \varepsilon.$$

Then,

$$0 \leq U(g, \mathcal{P}_{[a,b]}) - L(g, \mathcal{P}_{[a,b]}) < \varepsilon.$$

So,  $g$  is integrable as well. Moreover,

$$\int_a^b g(x) \, dx = \int_A f.$$

Q.E.D. ■

### Example 9.1.5

Compute the volume of the region

$$A = \{(x, y, z) \mid x \geq 0, y \geq 0, z \geq 0, x + y + z \leq 1\}$$

by integration.

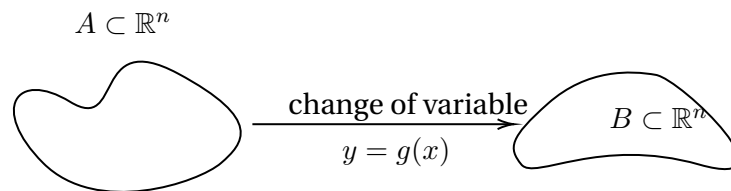
**Solution 3.**

$$v(A) \int_A \mathbb{1}_A = \int_0^1 \int_0^{1-x} \int_0^{1-x-y} 1 \, dz \, dy \, dx.$$

□

## 9.2 Change of Variable

**General Setting:**  $f : B \rightarrow \mathbb{R}$  bounded is an integrable function



**Goal:** Transform integral  $\int_B f(y)$  to an integral on  $A$ .

### Example 9.2.1 1D Case

$$\int f(y) \, dy = \int f(g(x)) \underbrace{g'(x) \, dx}_{dy}.$$

**Theorem 9.2.2 Change of Variable Formula in Higher Dimension**

Assume  $J_g(x) \neq 0 \quad \forall x \in A$ . If  $f : B \rightarrow \mathbb{R}$  is bounded and integrable on  $B = g(A)$ , then  $f \circ g(x) \cdot (J_g(x))$  is integrable on  $A$ , and

$$\int_B f(y) \, dy = \int_A f(g(x)) \cdot \underbrace{|J_g(x)|}_{dy} \, dx.$$

**Proof 1.** (Sketch)

- Change of volume under linear map:

Let  $\mathbf{L} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be a linear map given by

$$\mathbf{L} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}.$$

Denote  $y = \mathbf{L}x$ . Then,

$$v(L(S)) = \left| \det \begin{bmatrix} a & b \\ c & d \end{bmatrix} \right| \cdot v(S).$$

- Linear approximation of  $g : A \rightarrow B$ :

Fix  $x_0 \in A$ . Then, in a neighborhood of  $x_0$ ,  $g$  can be approximated by a linear map:

$$g(x) = g(x_0) + \mathbb{D}g(x_0)(x - x_0) + \text{error}.$$

- Conversion into integral formula:

Fix small rectangles  $S$  in  $A$ . Then,  $g(S)$  is “nearly” parallelogram. So,

$$v(g(S)) \approx |J_g(x_0)|v(S).$$

Do this for each rectangle  $S_{ij}$  in a partition:

$$v(g(S_{ij})) \approx |J_g(x_{ij})|v(S_{ij}).$$

Then,

$$\begin{aligned} f(y_{ij})v(g(S_{ij})) &\approx f(g(x_{ij}))|J_g(x_{ij})|v(S_{ij}) \\ \sum f(y_{ij})v(g(S_{ij})) &\approx \sum f(g(x_{ij}))|J_g(x_{ij})|v(S_{ij}) \end{aligned}$$

Through the summation and limit process:

$$\int_B f(y) \, dy = \int_A f(g(x))|J_g(x)| \, dx.$$



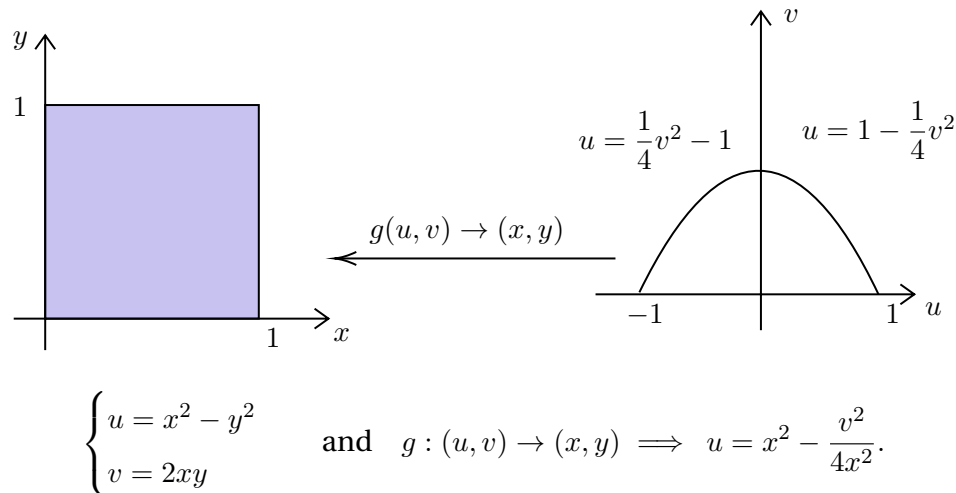
Q.E.D. ■

**Example 9.2.3**

Evaluate the integral using the change of variables  $u = x^2 - y^2$  and  $v = 2xy$ .

$$\int_0^1 \int_0^1 (x^2 + y^2) \sin(x^2 - y^2) \, dx \, dy.$$

1. Sketch the regions in  $xy$ -plane and  $uv$ -plane:



2. Compute the determinant:  $g^{-1} : (u, v) \rightarrow (x, y)$ .

$$J_{g^{-1}}(x, y) = \begin{vmatrix} \partial u / \partial x & \partial u / \partial y \\ \partial v / \partial x & \partial v / \partial y \end{vmatrix} = \begin{vmatrix} 2x & -2y \\ 2y & 2x \end{vmatrix} = 4x^2 + 4y^2.$$

So,

$$J_g = \frac{1}{J_{g^{-1}}(x, y)} = \frac{1}{4(x^2 + y^2)}$$

3. Apply the change of variable formula:

$$\begin{aligned} \int_0^1 \int_0^1 (x^2 + y^2) \sin(x^2 - y^2) \, dx \, dy &= \int_0^2 \int_{(1/4) \cdot v^2 - 1}^{1 - (1/4) \cdot v^2} (x^2 + y^2) \sin(x^2 - y^2) |J_g(x)| \, du \, dv \\ &= \int_0^2 \int_{(1/4) \cdot v^2 - 1}^{1 - (1/4) \cdot v^2} \cancel{(x^2 + y^2)} \sin(u) \frac{1}{4 \cancel{(x^2 + y^2)}} \, du \, dv \\ &= \frac{1}{4} \int_0^2 \int_{(1/4) \cdot v^2 - 1}^{1 - (1/4) \cdot v^2} \sin(u) \, du \, dv. \end{aligned}$$

**Remark 9.2 (Special Coordinate Systems)**

- *Polar Coordinate in  $\mathbb{R}^2$ :*

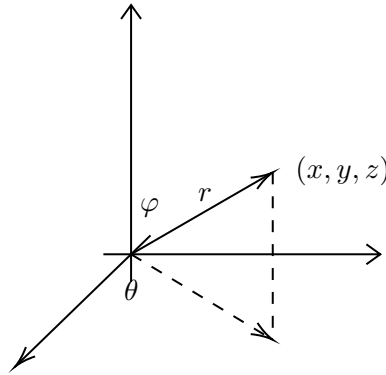
$$\begin{cases} x = r \cos \theta \\ y = r \sin \theta \end{cases} \quad J_g(r, \theta) = r,$$

$$\implies \int_B f(x, y) \, dx \, dy = \int_A f(r \cos \theta, r \sin \theta) r \, dr \, d\theta.$$

- *Spherical Coordinate in  $\mathbb{R}^3$ :*

$$\begin{cases} x = r \sin \varphi \cos \theta \\ y = r \sin \varphi \sin \theta \\ z = r \cos \varphi \end{cases} \quad J_g(r, \theta, \varphi) = r^2 \sin \varphi$$

$$\implies \int_B f(x, y, z) \, dx \, dy \, dz = \int_A f(r \sin \varphi \cos \theta, r \sin \varphi \sin \theta, r \cos \varphi) r^2 \sin \varphi \, dr \, d\theta \, d\varphi.$$



- *Cylindrical Coordinate in  $\mathbb{R}^3$ :*

$$\begin{cases} x = r \cos \theta \\ y = r \sin \theta \\ z = z \end{cases} \quad J_g(r, \theta, z) = r$$

$$\implies \int_B f(x, y, z) \, dx \, dy \, dz = \int_A f(r \cos \theta, r \sin \theta, z) r \, dr \, d\theta \, dz.$$

#### Example 9.2.4

- Evaluate  $\int_{-\infty}^{\infty} e^{-x^2} \, dx$

**Solution 2.**

Step 1 Evaluate integral

$$\int_{\mathbb{R}^2} e^{-x^2 - y^2} \, dx \, dy$$

by polar coordinate ( $x = r \cos \theta$  and  $y = r \sin \theta$ ). Let  $D_R$  denote the circle centered at origin with radius  $R$ . Then,

$$\begin{aligned} \int_{D_R} e^{-x^2-y^2} dx dy &= \int_0^{2\pi} \int_0^R e^{-r^2} r dr d\theta = \int_0^{2\pi} \left( -\frac{1}{2} e^{-r^2} \right) \Big|_0^R d\theta \\ &= 2\pi \left( -\frac{1}{2} e^{-R^2} + \frac{1}{2} \right) = -\pi e^{-R^2} + \pi. \end{aligned}$$

So,

$$\begin{aligned} \int_{\mathbb{R}^2} e^{-x^2-y^2} dx dy &= \lim_{R \rightarrow \infty} \int_{D_R} e^{-x^2-y^2} dx dy \\ &= \lim_{R \rightarrow \infty} \left( -\pi e^{-R^2} + \pi \right) \\ &= \pi. \end{aligned}$$

**Step 2** Evaluate

$$\int_{\mathbb{R}^2} e^{-x^2-y^2} dx dy$$

by Fubini's Theorem.

Let  $S_b = [-b, b] \times [-b, b] \subset \mathbb{R}^2$ . Then,

$$\begin{aligned} \int_{\mathbb{R}^2} e^{-x^2-y^2} dx dy &= \lim_{b \rightarrow \infty} \int_{S_b} e^{-x^2-y^2} dx dy \\ &= \lim_{b \rightarrow \infty} \int_{-b}^b \int_{-b}^b e^{-x^2} \cdot e^{-y^2} dx dy \\ &= \lim_{b \rightarrow \infty} \left( \int_{-b}^b e^{-x^2} dx \right) \cdot \left( \int_{-b}^b e^{-y^2} dy \right) \\ &= \left( \int_{-\infty}^{\infty} e^{-x^2} dx \right)^2 \end{aligned}$$

**Step 3** Combine Steps 1 and 2:

$$\pi = \int_{\mathbb{R}^2} e^{-x^2-y^2} dx dy = \left( \int_{-\infty}^{\infty} e^{-x^2} dx \right)^2$$

So,

$$\int_{-\infty}^{\infty} e^{-x^2} dx = \sqrt{\pi}.$$

□

- Evaluate  $\int_{\mathbb{R}^3} \frac{1}{x^2 + y^2 + z^2} dx dy dz$
- Evaluate  $\int_R 2e^{x^2-y^2} dx dy dz$ , where  $R = \{(x, y, z) \mid x^2 + y^2 \leq 1, 1 \leq x \leq 2\}$ .

--

## 10 Fourier Analysis

### 10.1 Introduction

**General Idea:** Try to decompose certain objects into simpler components.

- Algebraic Model:  $\mathbb{R}^n$

$$x = \sum_{i=1}^n x_i e_i,$$

where  $e_i$ 's are the standard basis.

- Calculus Model: Taylor Series

$$f(x) = \sum_{n=0}^{\infty} c_n (x - a)^n.$$

- Fourier Analysis: Theory of infinite dimensional inner product space of functions.

**Goal:** Decompose a function  $f(x)$  into a “linear combination of basis:”

$$f(x) = \sum_{n=-\infty}^{\infty} c_n \varphi_n(x).$$

**Physics Motivation:** Decompose complicated waves into harmonies.

### 10.2 Inner Product Space of Functions

#### 10.2.1 Basic Concepts

**Definition 10.2.1 (Inner Product).** Let  $V$  be a complex vector space. Then, an *inner product* on  $V$  is a map  $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{C}$  s.t.  $\forall f, g, h \in V$  and  $a, b \in \mathbb{C}$ , we have

- Linearity:

$$\langle af + bg, h \rangle = a \langle f, h \rangle + b \langle g, h \rangle.$$

- Conjugate Symmetry:

$$\langle f, g \rangle = \overline{\langle g, f \rangle}.$$

- Positive Definiteness:

$$\langle f, f \rangle \geq 0 \quad \text{and} \quad \langle f, f \rangle = 0 \iff f = 0.$$

#### Example 10.2.2

$\mathbb{C}$  is an inner product space under the inner product:

$$\langle z_1, z_2 \rangle = z_1 \overline{z_2}.$$

**Corollary 10.2.3 Conjugate Linearity in the Second Component:**

$$\langle h, af + bg \rangle = \bar{a} \langle h, f \rangle + \bar{b} \langle h, g \rangle.$$

**Proof 1.**

$$\begin{aligned} \langle h, af + bg \rangle &= \overline{\langle af + bg, h \rangle} && \text{[Conjugate symmetry]} \\ &= \overline{a \langle f, g \rangle + b \langle g, h \rangle} && \text{[Linearity]} \\ &= \bar{a} \langle h, f \rangle + \bar{b} \langle h, g \rangle. && \text{[Conjugate symmetry]} \end{aligned}$$

Q.E.D. ■

**Definition 10.2.4 (Norm and Distance Induced by Inner Product).**

- Norm:

$$\|f\| := \sqrt{\langle f, f \rangle}.$$

- Distance from  $f$  to  $g$ :

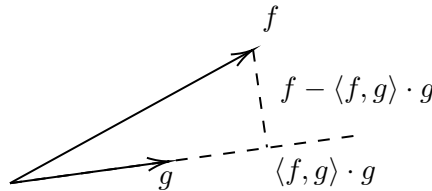
$$d(f, g) := \|f - g\|.$$

**Corollary 10.2.5 Facts:**

- $(V, \|\cdot\|)$  is a normed space.
- $(V, d)$  is a metric space.

**Lemma 10.2.6 Cauchy-Schwarz Inequality:**

$$|\langle f, g \rangle| \leq \|f\| \cdot \|g\|$$

**Proof 2.**

The projection should have the smallest length:

$$\begin{aligned} 0 &\leq \|f - \langle f, g \rangle g\|^2 = \langle f, \langle f, g \rangle g, f - \langle f, g \rangle g \rangle \\ &= \langle f, f - \langle f, g \rangle g \rangle - \langle f, g \rangle \langle g, f - \langle f, g \rangle g \rangle \\ &= \langle f, f \rangle - \overline{\langle f, g \rangle} \langle f, g \rangle - \langle f, g \rangle \langle g, f \rangle + \langle f, g \rangle \overline{\langle f, g \rangle} \langle g, g \rangle \\ &= \|f\|^2 - |\langle f, g \rangle|^2 - |\langle f, g \rangle|^2 + |\langle f, g \rangle|^2 \|g\|^2. \end{aligned}$$

Normalize: let  $\|g\| = 1$ . Then,

$$\begin{aligned} 0 &\leq \|f\|^2 - |\langle f, g \rangle|^2 \\ |\langle f, g \rangle|^2 &\leq \|f\|^2 \\ |\langle f, g \rangle| &\leq \|f\| = \|f\| \cdot \|g\|. \end{aligned}$$

Q.E.D. ■

**Definition 10.2.7 (Convergence).** Suppose  $f_n, f \in V$ . Then,  $f_n \rightarrow f$  in  $V$  if  $\|f_n - f\| \rightarrow 0$  as  $n \rightarrow \infty$ . We call this *convergence in norm*.

### 10.2.2 The Space $\mathcal{C}$ and $L^2$

**Definition 10.2.8 (Integral of Complex Valued Functions).** Suppose  $f(x) = f_1(x) + \mathbf{i}f_2(x) : [a, b] \rightarrow \mathbb{C}$  be a complex-valued function, where  $f_1, f_2 : [a, b] \rightarrow \mathbb{R}$ . Then,

$$\int_a^b f(x) \, dx := \int_a^b f_1(x) \, dx + \mathbf{i} \int_a^b f_2(x) \, dx.$$

**Definition 10.2.9 (The Space  $\mathcal{C}$  and  $L^2$ ).** Fix an interval  $[a, b]$ .

- $\mathcal{C} := \{f(x) \mid f : [a, b] \rightarrow \mathbb{C} \text{ is continuous}\}.$
- $L^2 := \left\{ f : [a, b] \rightarrow \mathbb{C} \mid \int_a^b |f(x)|^2 \, dx < \infty \right\}.$

The condition  $\int_a^b |f(x)|^2 \, dx < \infty$  is called  $L^2$  *integrable*.

**Corollary 10.2.10 Facts:**

- $\mathcal{C}$  and  $L^2$  are vector spaces.  $\mathcal{C}$  is a subspace of  $L^2$ .
- Zero vector in  $\mathcal{C}$ :  $f(x) \equiv 0$ .
- Zero vector in  $L^2$ :  $f(x) = 0$  a. e. (almost everywhere).

That is,  $m(\{x \in [a, b] \mid f(x) \neq 0\}) = 0$ .

- $\underbrace{f_1 = f_2}_{\text{vectors}} \text{ in } L^2 \iff \underbrace{f_1(x) = f_2(x)}_{\text{function}} \text{ a. e.}$

- Inner Product:

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

**Claim 10.2.11** With the above definition of inner product,  $\mathcal{C}$  and  $L^2$  are inner product spaces.

### 10.3 Fourier Analysis on Inner Product Space

#### 10.3.1 Geometry of an Inner Product Space

**Definition 10.3.1 (Orthogonality).**  $f, g \in V$  are *orthogonal* (denoted as  $f \perp g$ ) if  $\langle f, g \rangle = 0$ .

**Definition 10.3.2 (Orthonormal Family).** A family  $\{\varphi_1, \varphi_2, \dots\} \subset V$  is called an *orthonormal family* if

- $\langle \varphi_i, \varphi_j \rangle = 0 \quad \forall i \neq j$
- $\text{norm} \varphi_i = 1 \quad \forall i$ .

Or equivalently,

$$\langle \varphi_i, \varphi_j \rangle = \delta_{ij} = \begin{cases} 1, & \text{if } i = j \\ 0, & \text{if } i \neq j. \end{cases}$$

#### Example 10.3.3

In  $\mathbb{R}^n$ :  $\{e_1, e_2, \dots, e_n\}$ , the standard basis, is an orthonormal basis.

#### Theorem 10.3.4 Gram-Schmidt Process: Generate Orthonormal Family from Linear Independent Family

$$\underbrace{\{g_0, g_1, \dots\}}_{\text{linear independent}} \rightarrow \underbrace{\{\varphi_0, \dots, \varphi_1, \dots\}}_{\text{orthonormal}}$$

1. Orthogonal projection:

$$x = \sum_i c_i e_i,$$

where  $c_i = \langle x, e_i \rangle$ . Then, we have

$$\langle x - \langle x, e_i \rangle e_i, e_i \rangle = 0.$$

2. Inductive Process:

$$\varphi_0 = \frac{g_0}{\|g_0\|}$$

$$f_1 = g_1 - \langle g_1, \varphi_0 \rangle \varphi_0, \quad \implies \varphi_1 = \frac{f_1}{\|f_1\|}$$

$\vdots$

$$f_n = g_n - \sum_{i=0}^{n-1} \langle g_n, \varphi_i \rangle \varphi_i, \quad \implies \varphi_n = \frac{f_n}{\|f_n\|}.$$



### 10.3.2 Fourier Series and Complete Family

**Definition 10.3.5 (Complete Orthonormal Family).** An orthonormal family  $\{\varphi_0, \varphi_1, \dots\}$  (countable) is called *complete* if each  $f \in V$  can be written as

$$f = \sum_{k=0}^{\infty} c_k \varphi_k \quad (\star)$$

**Remark 10.1**

- The meaning of  $(\star)$ :

$$\left\| f - \sum_{k=0}^n c_k \varphi_k \right\| \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

- $(\star)$  is called the Fourier series of  $f$  w.r.t.  $\{\varphi_0, \varphi_1, \dots\}$ .
- If  $\{\varphi_0, \varphi_1, \dots\}$  is complete, then it is an orthonormal basis of  $V$ .

**Objective:** Find suitable complete orthonormal family and expand  $f \in V$  into Fourier series.

**Theorem 10.3.6**

If  $f$  has Fourier series expansion:

$$f = \sum_{k=0}^{\infty} c_k \varphi_k,$$

then,

$$c_k = \langle f, \varphi_k \rangle \quad \text{for } k = 0, 1, \dots$$

$c_k$ 's are called the *Fourier coefficients* of  $f$ .

**Proof 1.** Let

$$S_n = \sum_{k=0}^n c_k \varphi_k.$$

Then,

$$\|f - S_n\| \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Fix  $m \geq 0$ . Then, for any  $n \geq m$ ,

$$\begin{aligned} c_m &\stackrel{\text{want}}{=} \langle f, \varphi_m \rangle = \langle f - S_n + S_n, \varphi_m \rangle \\ &= \langle f - S_n, \varphi_m \rangle + \langle S_n, \varphi_m \rangle \\ &= \langle f - S_n, \varphi_m \rangle + c_m \\ &= 0 + c_m \quad \text{as } n \rightarrow \infty \end{aligned}$$

[orthogonality]

[Cauchy-Schwarz]  
 $\|f - S_n\| \rightarrow 0$

So,  $\langle f, \varphi_m \rangle = c_m$ .

Q.E.D. ■

**Question:** Given  $f$  and  $\{\varphi_1, \varphi_2, \dots\}$ , does the series

$$\sum_{k=0}^{\infty} \langle f, \varphi_k \rangle \varphi_k$$

converge to  $f$ ?

**Theorem 10.3.7 Properties of Fourier Coefficients**

Assume  $\{\varphi_0, \varphi_1, \dots\}$  is an orthonormal family in  $V$ .

- Bessel's Inequality:

$$\sum_{k=0}^{\infty} |\langle f, \varphi_k \rangle|^2 \leq \|f\|^2.$$

- Parseval's Equality (One can View this as the Pythagorean Theorem):

If

$$f = \sum_{k=0}^{\infty} \langle f, \varphi_k \rangle \varphi_k,$$

then

$$\sum_{k=0}^{\infty} |\langle f, \varphi_j \rangle|^2 = \|f\|^2.$$

**Proof2.** Let  $S_n = \sum_{k=0}^n \langle f, \varphi_k \rangle \varphi_k$ . Denote  $c_k = \langle f, \varphi_k \rangle$ .

$$\begin{aligned} \|f\|^2 &= \|f - S_n + S_n\|^2 \\ &= \langle f - S_n + S_n, f - S_n + S_n \rangle && \text{[definition]} \\ &= \|f - S_n\|^2 + \|S_n\|^2 && \text{[Linearity, } f - S_n \perp S_n] \\ \|S_n\| &= \langle S_n, S_n \rangle = \sum_{k=0}^n |c_k|^2. \end{aligned}$$

Then,

$$\|f\|^2 = \underbrace{\|f - S_n\|^2}_{\geq 0} + \sum_{k=0}^n |c_k|^2 \implies \|f\|^2 \geq \sum_{k=0}^n |c_k|^2 = \sum_{k=0}^n |\langle f, \varphi_k \rangle|^2$$

true for any  $n$ . So, we get ① by letting  $n \rightarrow \infty$ .

Under the assumption of ②, when  $n \rightarrow \infty$ , we have  $\|f - S_n\|^2 \rightarrow 0$ . So,

$$\|f\|^2 = \sum_{k=0}^{\infty} |c_k|^2 = \sum_{k=0}^{\infty} |\langle f, \varphi_k \rangle|^2.$$

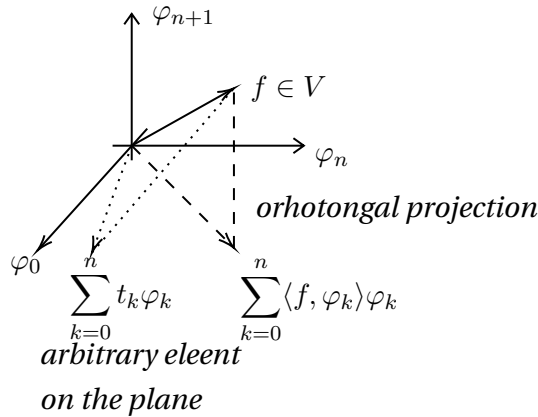
Q.E.D. ■

**Theorem 10.3.8 Best mean Approximation Theorem (BMAT)**

Assume  $\{\varphi_0, \varphi_1, \dots\}$  is an orthonormal family in  $V$ . For any scalars  $t_0, t_1, \dots, t_n \in \mathbb{C}$ , we have

$$\left\| f - \sum_{k=0}^n t_k \varphi_k \right\| \geq \left\| f - \sum_{k=0}^n \langle f, \varphi_k \rangle \varphi_k \right\|.$$

- The first sum is an arbitrary element in the plane formed by  $\{\varphi_0, \dots, \varphi_n\}$ .
- The second sum is the orthogonal projection of  $f$  onto the plane.

**Remark 10.2 (Geometric Interpretation)**

$LHS \leq RHS$ : the shortest distance from a point  $f$  to the plane is achieved by the orthogonal projection (or, the perpendicular line).

**Proof 3.** Let  $h_n = \sum_{k=0}^n t_k \varphi_k$ . Then,

$$\begin{aligned} \|f - h_n\|^2 &= \langle f - h_n, f - h_n \rangle \\ &= \langle f, f \rangle - \langle h_n, f \rangle - \langle f, h_n \rangle + \langle h_n, h_n \rangle \\ &= \|f\|^2 - \sum_{k=0}^n t_k \overline{c_k} - \sum_{k=0}^n \overline{t_k} c_k + \sum_{k=0}^n |t_k|^2 \\ &\quad \vdots \text{ linearity} \\ &= \|f\|^2 - \sum_{k=0}^n |c_k|^2 + \sum_{k=0}^n |t_k - c_k|^2 \\ &= \|f - f_n\|^2 + \underbrace{\sum_{k=0}^n |t_k - c_k|^2}_{\geq 0}. \end{aligned}$$

So, BMAT is proven.

Q.E.D. ■

## 10.4 Completeness and Convergence in $L^2$

### Theorem 10.4.1 Orthogonal Functions in $L^2$

Let  $V = L^2([a, b])$ , where  $[a, b] = [0, 2\pi]$ .

- Exponential family:

$$\varphi_n(x) = \frac{e^{inx}}{\sqrt{2\pi}}, \quad n = 0, \pm 1, \pm 2, \dots$$

- Trig. family:

$$\frac{1}{\sqrt{2\pi}}, \frac{\cos mx}{\sqrt{2\pi}}, \frac{\sin nx}{\sqrt{2\pi}}, \quad n, m = 1, 2, \dots$$

**Claim 10.4.2** Both families are orthogonal.

**Proof 1.** (of exponential family)

WTS:

$$\langle \varphi_n, \varphi_m \rangle = \delta_{n,m} = \begin{cases} 1, & n = m \\ 0, & n \neq m \end{cases}$$

$$\begin{aligned} \langle \varphi_n, \varphi_m \rangle &= \int_0^{2\pi} \varphi_n(x) \overline{\varphi_m(x)} dx \\ &= \frac{1}{2\pi} \int_0^{2\pi} e^{inx} \cdot e^{-imx} dx \\ &= \frac{1}{2\pi} \int_0^{2\pi} e^{i(n-m)x} dx \\ &= \begin{cases} 1, & n = m \\ \frac{1}{2\pi} \cdot \frac{1}{i(n-m)} e^{i(n-m)x} \Big|_0^{2\pi} = 0, & n \neq m. \end{cases} \end{aligned}$$

Q.E.D. ■

### Theorem 10.4.3 Mean Convergence Property/Completeness

The exponential family  $\{\varphi_n\}_{n=-\infty}^{\infty}$  is complete in  $L^2$

**Remark 10.3** To prove this Theorem, we aim to show: any function  $f(x) \in L^2$  can be represented by its Fourier series:

$$f(x) = \sum_{n=-\infty}^{\infty} \langle f, \varphi_n \rangle \varphi_n.$$

i.e.,

$$\left\| f(x) - \sum_{k=-n}^n \langle f, \varphi_k \rangle \varphi_k \right\|_{L^2} \xrightarrow{(n \rightarrow \infty)} 0.$$

**Lemma 10.4.4 Stone-Weierstrass Theorem:** Continuous functions can be approximated by polynomials of  $e^{ix}$  and  $e^{-ix}$ . More precisely, given  $f : [0, 2\pi] \rightarrow \mathbb{C}$  continuous with  $f(0) = f(2\pi)$ . Then,  $\forall \varepsilon > 0$ ,

$\exists n \geq 1$  and  $c_k, k = 0, \pm 1, \dots$  s.t.

$$|f(x) - p_n(x)| < \varepsilon \quad \forall x \in [0, 2\pi],$$

where

$$p_n(x) = \sum_{k=-n}^n c_k e^{ikx},$$

a polynomial in  $e^{ix}$  and  $e^{-ix}$ .

**Lemma 10.4.5:** Integrable functions can be approximated by continuous functions. That is, let  $f \in L^2$  and  $\varepsilon > 0$  be given,  $\exists$  continuous function  $g : [0, 2\pi] \rightarrow \mathbb{C}$  with  $g(0) = g(2\pi)$  s.t.

$$\|f - g\| < \varepsilon.$$

### ► Proof 2 of Mean Convergence Property

- Step 1 Special Case:

Let  $f$  be continuous with  $f(0) = f(2\pi)$ . Write

$$S_n = \sum_{k=-n}^n \langle f, \varphi_k \rangle \varphi_k, \quad \text{where } \varphi_k(x) = \frac{e^{ikx}}{\sqrt{2\pi}}$$

WTS:  $\|f - S_n\| \rightarrow 0$  as  $n \rightarrow \infty$ .

Fix  $\varepsilon > 0$ . By Lemma 10.4.4, we can choose  $p_N(x)$  s.t.

$$|f(x) - p_N(x)| < \frac{\varepsilon}{\sqrt{2\pi}} \quad \forall x \in [0, 2\pi].$$

Then,

$$\|f - p_N\| \leq \left( \int_0^{2\pi} \left( \frac{\varepsilon}{\sqrt{2\pi}} \right)^2 \right)^{1/2} = \varepsilon.$$

Thus,  $\forall n \geq N$ , we have

$$\begin{aligned} \|f - S_n\| &\leq \|f - S_N\| && [\text{BMAT. } L_N \subset L_n \implies S_N \in L_n.] \\ &\leq \|f - p_N\| && [\text{BMAT. } p_N \in L_N] \\ &\leq \varepsilon. \end{aligned}$$

So,  $\|f - S_n\| \rightarrow 0$  as  $n \rightarrow \infty$ .

- Step 2 General Case:

Fix  $f \in L^2$ . WTS:  $f = \sum_{k=-\infty}^{\infty} \langle f, \varphi_k \rangle \varphi_k$ .

By Lemma 10.4.5,  $\exists$  sequence of continuous functions  $g_n : [0, 2\pi] \rightarrow \mathbb{C}$  with  $g(0) = g(2\pi)$  s.t.

$$\|f - g_n\| \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

By Step 1, for each  $g_n$ , we have

$$g_n = \sum_{k=-\infty}^{\infty} \langle g_n, \varphi_k \rangle \varphi_k.$$

WTS:  $\|f - S_n\| \rightarrow 0$ .

Fix  $\varepsilon > 0$ . Choose  $N$  s.t.

$$\|f - g_N\| < \frac{\varepsilon}{3}.$$

Then, choose  $M$  s.t.

$$n \geq M \implies \|g_N - S_n(g_N)\| < \frac{\varepsilon}{3},$$

where  $S_n(g_N)$  denotes the partial sum of Fourier series of  $g_N$ .

$$S_n(g_N) = \sum_{k=-n}^n \langle g_N, \varphi_k \rangle \varphi_k.$$

Thus,  $\forall n \geq M$ , we have

$$\begin{aligned} \|f - S_n\| &= \|S_n - S_n(g_N) + S_n(g_N) - g_N + g_N - f\| \\ &\leq \|S_n - S_n(g_N)\| + \|S_n(g_N) - g_N\| + \|g_N - f\| \\ \|S_n - S_n(g_N)\| &= \left\| \sum_{k=-n}^n \langle f, \varphi_k \rangle \varphi_k - \sum_{k=-n}^n \langle g_N, \varphi_k \rangle \varphi_k \right\| \\ &= \left\| \sum_{k=-n}^n \langle f - g_N, \varphi_k \rangle \varphi_k \right\| \\ &= \left\langle \sum_{k=-n}^n \langle f - g_N, \varphi_k \rangle \varphi_k, \sum_{k=-n}^n \langle f - g_N, \varphi_k \rangle \varphi_k \right\rangle^{1/2} \\ &= \left( \sum_{k=-n}^n |\langle f - g_N, \varphi_k \rangle|^2 \right)^{1/2} \quad \text{[Pythagorean Theorem]} \\ &\leq \|f - g_N\| < \frac{\varepsilon}{3}. \end{aligned}$$

So,

$$n \geq M \implies \|f - S_n\| < \frac{\varepsilon}{3} + \frac{\varepsilon}{3} + \frac{\varepsilon}{3} = \varepsilon.$$

Therefore,

$$\|f - S_n\| \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Q.E.D. ■

---

*With that, these notes mark the end of a journey through the rigorous landscapes of Real Analysis. From the foundational structure of  $\mathbb{R}$  to the elegance of Fourier series in  $L^2$ , this document reflects not only the theorems and proofs, but also the quiet persistence of curiosity.*

*I would like to express my sincere gratitude to:*

***Professor Shanshuang Yang***

*for his guidance, clarity, and intellectual generosity throughout MATH 411 & 412.*

***Jerrold E. Marsden and Michael J. Hoffman***

*whose text Elementary Classical Analysis served as a source of inspiration, challenge, and reflection.*

***My peers and friends***

*for the many shared ideas, problem-solving moments, and philosophical tangents.*

***Myself***

*for persistence, patience, and the willingness to wrestle with the abstract.*

***End of Notes***

---

*Jiuru Lyu*

*June 18, 2025*