PMATH351 - Real Analysis

CLASSNOTES FOR FALL 2018

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1 Lecture 1 Sep 06th

1.1 Course Logistics

No content is covered in today's lecture so this chapter will cover some of the important logistical highlights that were mentioned in class.

- Assignments are designed to help students understand the content.
- Due to shortage of manpower, not all assignment questions will be graded; however, students are encouraged to attempt all of the questions.
- To further motivate students to work on ungraded questions, the midterm and final exam will likely recycle some of the assignment questions.
- There are no required text, but the professor has prepared course notes for reading. The course note are self-contained.
- The approach of the class will be more interactive than most math courses.
- Due to the size of the class, students are encouraged to utilize Waterloo Learn for questions, so that similar questions by multiple students can be addressed at the same time.

1.2 *Preview into the Introduction*

How do we compare the size of two sets?

- If the sets are finite, this is a relatively easy task.
- If the sets are infinite, we will have to rely on functions.

- Injective functions tell us that the domain is of size that is lesser than or equal to the codomain.
- Surjective functions tell us that the codomain is of size that is lesser than or equal to the domain.
- So does a bijective function tell us that the domain and codomain have the same size? Yes, although this is not as intuitive as it looks, as it relies on Cantor-Schröder-Bernstein Theorem.

Now, given two arbitrary sets, are we guaranteed to always be able to compare their sizes? It would be very tempting to immediately say yes, but to do that, one would have to agree on the Axiom of Choice. Fortunately, within the realm of this course, the Axiom of Choice is taken for granted.

2 Lecture 2 Sep 10th

2.1 Basic Set Theory

We shall use the following notations for some of the common set of numbers that we are already familiar with:

- N denotes the set of natural numbers {1,2,3,...};
- \mathbb{Z} denotes the set of integers {..., -2, -1, 0, 1, 2, ...};
- Q denotes the set of rational numbers $\left\{\frac{a}{b} \mid a \in \mathbb{Z}, b \in \mathbb{N}\right\}$; and
- \mathbb{R} denotes the set of real numbers.

We shall start with having certain basic properties of \mathbb{N} , \mathbb{Z} , and \mathbb{Q} .

WE WILL USE the notation $A \subset B$ and $A \subseteq B$ interchangably to mean that A is a subset of B with the possibility that A = B. When we wish to explicitly emphasize this possibility, we shall use $A \subseteq B$. When we wish to explicitly state that A is a **proper subset** of B, we will either specify that $A \neq B$ or simply $A \subseteq B$.

Definition 1 (Universal Set)

A universal set, which we shall generally give the label X, is a set that contains all the mathematical objects that we are interested in.

With a universal set in place, we can have the following definitions:

Definition 2 (Union)

This is a hand-wavy definition, but it is not in the interest of this course to further explore on this topic.

Let X be a set. If $\{A_{\alpha}\}_{{\alpha}\in I}$ such that $A_{\alpha}\subset X$, then the union for all A_{α} is defined as

$$\bigcup_{\alpha\in I}A_{\alpha}:=\{x\in X\mid \exists \alpha\in I, x\in A_{\alpha}\}.$$

Definition 3 (Intersection)

Let X be a set. If $\{A_{\alpha}\}_{{\alpha}\in I}$ such that $A_{\alpha}\subset X$, then the **intersection** for all A_{α} is defined as

$$\bigcap_{\alpha\in I}A_{\alpha}:=\{x\in X\mid \forall \alpha\in I, x\in A_{\alpha}\}.$$

Definition 4 (Set Difference)

Let X be a set and A, B \subseteq X. The **set difference** of A from B is defined as

$$A \setminus B := \{ x \in X \mid x \in A, x \notin B \}.$$

On a similar notion:

Definition 5 (Symmetric Difference)

Let X be a set and A, B \subseteq X. The **symmetric difference** of A and B is defined as

$$A\Delta B := \{ x \in X \mid (x \in A \land x \notin B) \lor (x \notin A \land x \in B) \}.$$

We can also talk about the non-members of a set:

In words, for an element in the symmetric difference of two sets, the element is either in A or B but not both. We can also think of the symmetric difference

$$(A \cup B) \setminus (A \cap B)$$

or

 $(A \setminus B) \cup (B \setminus A).$

Definition 6 (Set Complement)

Let X be a set and $A \subset X$. The set of all non-members of A is called the **complement** of A, which we denote as

$$A^c := \{ x \in X \mid x \notin A \}.$$

66 Note

Note that

$$(A^c)^c = \{x \in X \mid x \notin A^c\} = \{x \in X \mid x \in A\} = A.$$

Now taking a step away from that, we define the following:

Definition 7 (Empty Set)

An *empty set*, denoted by \emptyset , is a set that contains nothing.

66 Note

The empty set is set to be a subset of all sets.

Definition 8 (Power Set)

Let X be a set. The power set of X is the set that contains all subsets of X,

$$\mathcal{P}(X) := \{ A \mid A \subset X \}.$$

66 Note

A power set is always non-empty, since $\emptyset \in \mathcal{P}(\emptyset)$ *, and since* $\emptyset \subset X$ *for* any set X, we have $\emptyset \in \mathcal{P}(X)$.

Example 2.1.1

Let $X = \{1, 2, ..., n\}$. There are several ways we can show that the size of $\mathcal{P}(X)$ is 2^n . One of the methods is by using a characteristic function that maps from A to $\{0,1\}$, defined by

$$X_A: A \to \{0,1\}$$

$$X_A(x) = \begin{cases} 1 & x \in A \\ 0 & x \notin A \end{cases}.$$

Using this function, each element in X have 2 states: one being in the subset, and the other being not in the subset, which are represented by 1 and 0 respectively. It is then clear that there are 2^n of such configurations.

Theorem 1 (De Morgan's Laws)

Let X be a set. Given $\{A_{\alpha}\}_{{\alpha}\in I}\subset \mathcal{P}(X)$, we have

1.
$$\left(\bigcup_{\alpha\in I}A_{\alpha}\right)^{c}=\bigcap_{\alpha\in I}A_{\alpha}^{c}$$
; and

2.
$$\left(\bigcap_{lpha\in I}A_lpha
ight)^c=igcup_{lpha\in I}A_lpha^c$$
.

Proof

1. Note that

$$x \in \left(\bigcup_{\alpha \in I} A_{\alpha}\right)^{c} \iff \nexists \alpha \in I \ x \in A_{\alpha}$$

$$\iff \forall \alpha \in I \ x \notin A_{\alpha}$$

$$\iff \forall \alpha \in I \ x \in A_{\alpha}^{c} \text{ by set complementation}$$

$$\iff x \in \bigcap_{\alpha \in I} A_{\alpha}^{c}.$$

2. Observe that, by part 1,

$$\left(\bigcap_{\alpha\in I}A_{\alpha}\right)^{c}=\left(\left(\bigcup_{\alpha\in I}A_{\alpha}^{c}\right)^{c}\right)^{c}=\bigcup_{\alpha\in I}A_{\alpha}^{c}.$$

Suppose $I=\emptyset$. Then what is $\bigcup_{\alpha\in\emptyset}A_{\alpha}$? It is sensible to think that all we are left with is simply a union of empty sets, and so

$$\bigcup_{\alpha\in\emptyset}A_{\alpha}=\emptyset. \tag{2.1}$$

And what about $\bigcap A_{\alpha}$? By \blacksquare Theorem 1, it is quite clear from Equation (2.1) that

$$\bigcap_{\alpha\in\emptyset}A_{\alpha}=X.$$

2.2 Products of Sets

Definition 9 (Product of Sets)

Given 2 sets X and Y, the **product** of X and Y is given by

$$X \times Y := \{(x,y) \mid x \in X, y \in Y\}.$$

We often refer to elements of $X \times Y$ as **tuples**.

66 Note

Now if

$$X = \{x_1, x_2, ..., x_n\},\$$

 $Y = \{y_1, y_2, ..., y_m\},\$

then

$$X \times Y = \{(x_i, y_i) \mid i = 1, 2, ..., n, j = 1, 2, ..., m\}$$

and so the size of $X \times Y$ is mn.

Consequently, we can think of tuples as two elements being in some "relation".

Definition 10 (Relation)

A **relation** on sets X and Y is a subset R of the product $X \times Y$. We write

$$xRy$$
 if $(x,y) \in R \subset X \times Y$.

We call

• $\{x \in X \mid \exists y \in Y, (x,y) \in R\}$ as the domain of R; and

• $\{y \in Y \mid \exists x \in X, (x,y) \in R\}$ as the range of R.

In relation to that, functions are, essentially, relations.

Definition 11 (Function)

A function from X to Y is a relation R such that

$$\forall x \in X \exists ! y \in Y (x, y) \in R.$$

Suppose $X_1, X_2, ..., X_n$ are non-empty¹ sets. We can define

$$X_1 \times X_2 \times ... \times X_n = \prod_{i=1}^n X_i := \{(x_1, x_2, ..., x_n) \mid x_i \in X_i\}.$$

Now if $X_i = X_j = X$ for all i, j = 1, 2, ..., n, we write

$$\prod_{i=1}^n X_i = \prod_{i=1}^n X = X^n.$$

And now comes the problem: given a collection $\{X_{\alpha}\}_{\alpha\in I}$ of non-empty sets², what do we mean by

$$\prod_{\alpha \in I} X_{\alpha}?$$

To motivate for what comes next, consider

$$\prod_{i=1}^{n} X_{i} = X_{1} \times \ldots \times X_{n} = \{(x_{1}, ..., x_{n}) \mid x_{i} \in X_{i}\}.$$

Choose $(x_1,...,x_n) \in \prod_{i=1}^n X_i$. This induces a function

$$f_{(x_1,...,x_n)}: \{1,...,n\} \to \bigcup_{i=1}^n X_i$$

¹ We are typically only interested in non-empty sets, since empty sets usually lead us to vacuous truths, which are not interesting.

² i.e. we now talk about arbitrary $\alpha \in I$.

with

$$f(1) = x_1 \in X_1$$

$$f(2) = x_2 \in X_2$$

$$\vdots$$

$$f(n) = x_n \in X_n$$

Now assume for a more general *f* such that

$$f:\{1,...,n\}\to \bigcup_{i=1}^n X_i$$

is defined by

$$f(i) \in X_i$$
.

Then, we have

$$(f(1), f(2), ..., f(n)) \in \prod_{i=1}^{n} X_i,$$

which leads us to the following notion:

Definition 12 (Choice Function)

Given a collection $\{X_{\alpha}\}_{{\alpha}\in I}$ of non-empty sets, let

$$\prod_{\alpha \in I} X_{\alpha} = \left\{ f : I \to \bigcup_{\alpha \in I} X_{\alpha} \right\}$$

such that $f(\alpha) \in X_{\alpha}$. Such an f is called a choice function.

And so we may ask a similar question as before: if each X_{α} is nonempty, is $\prod_{\alpha \in I} X_{\alpha}$ non-empty? Turns out this is not as easy to show. In fact, it is essentially impossible to show, because this is exactly the Axiom of Choice.

3 Lecture 3 Sep 12th

3.1 Axiom of Choice

Recall our final question of last lecture: If $\{X_{\alpha}\}_{\alpha \in I}$ is a non-empty collection of non-empty sets, is

$$\prod_{\alpha\in I}X_{\alpha}\neq\emptyset$$
?

Turns out this is widely known (in the world of mathematics) as the Axiom of Choice.

■ Axiom 2 (Zermelo's Axiom of Choice)

If $\{X_{\alpha}\}_{\alpha \in I}$ is a non-empty collection of non-empty sets, then

$$\prod_{\alpha\in I}X_{\alpha}\neq\emptyset.$$

An equivalent statement of the above axiom is:

▼ Axiom 3 (Zermelo's Axiom of Choice v2)

$$X \neq \emptyset \implies$$

$$\exists f: \mathcal{P}(X) \setminus \{\emptyset\} \to X \ \forall A \in \mathcal{P}(X) \setminus \{\emptyset\} \ f(A) \in A$$

where f is the choice function.

Exercise 3.1.1

Prove that \mathbf{U} *Axiom 2 and* \mathbf{U} *Axiom 3 are equivalent.*

Proof

From **U** Axiom 2 to **U** Axiom 3:

Since $X \neq \emptyset$, we have that $\mathcal{P}(X) \setminus \{\emptyset\}$ is a non-empty collection of non-empty sets. Therefore,

$$\prod_{A\in\mathcal{P}(X)\setminus\{\emptyset\}}A\neq\emptyset.$$

So we know that

$$\exists (x_A)_{A \in \mathcal{P}(X) \setminus \{\emptyset\}} \in \prod_{A \in \mathcal{P}(X) \setminus \{\emptyset\}} A.$$

We then simply need to choose the choice function $f: \mathcal{P}(X) \setminus \{\emptyset\} \to X$ such that

$$f(A) = x_A \in A$$
.

From **□** Axiom 3 to **□** Axiom 2:

Let $X_{\alpha} \in \mathcal{P}(X)$ for $\alpha \in I$, where I is some index set. We know that not all $X_{\alpha} = \emptyset$ since $X \neq \emptyset$. Choose $J \subseteq I$ such that $\{X_{\alpha}\}_{\alpha \in J}$ is a non-empty collection of non-empty sets. Let $f : \mathcal{P}(X) \setminus \{\emptyset\}$ be any choice function. By \P Axiom 3,

$$\forall X_{\alpha} \in \mathcal{P}(X) \setminus \{\emptyset\} \quad f(X_{\alpha}) \in X_{\alpha}.$$

Therefore,

$$(f(X_{\alpha}))_{\alpha\in J}\in\prod_{\alpha\in J}X_{\alpha}.$$

3.2 Relations

Now, it is in our interest to start talking about comparisons or relations between the mathematical objects that we have defined.

Definition 13 (Relations)

A relation R on a set X is 1

- (*Reflexive*) $\forall x \in X \ xRx$;
- (Symmetric) $\forall x, y \in X \ xRy \iff yRx$;
- (Anti-symmetric) $\forall x, y \in X \ xRy \land yRx \implies x = y$;
- (Transitive) $\forall x, y, z \in X \ xRy \land yRz \implies xRz$.

Example 3.2.1

Let $X = \mathbb{R}$, and let $xRy \iff x \le y$, where \le is the notion of "less than or equal to", which we shall assume that it has the meaning that we know. Observe that \leq is:

- reflexive: $\forall x \in \mathbb{R} \ x \leq x$ is true;
- anti-symmetric: $\forall x, y \in \mathbb{R} \ x \leq y \land y \leq x \implies x = y$; and
- transitive: $\forall x, y, z \in \mathbb{R} \ x \le y \land y \le z \implies x \le z$.

Example 3.2.2

Let $Y \neq \emptyset$, $X = \mathcal{P}(Y)$, with $ARB \iff A \subseteq B$. Observe that \subseteq is:

- reflexive: $\forall A \in \mathcal{P}(Y) \ ARA \iff A \subseteq A \text{ is true};$
- anti-symmetric: $\forall A, B \in \mathcal{P}(Y) \ ARB \land BRA \iff A \subseteq B \land B \subseteq$ $A \implies A = B;$
- transitive: $\forall A, B, C \in \mathcal{P}(Y) \ ARB \land BRC \iff A \subseteq B \land B \subseteq C \implies$ $A \subseteq C$.

Example 3.2.3

Let $Y \neq \emptyset$, $X = \mathcal{P}(Y)$, with $ARB \iff A \supseteq B$. Observe that \supseteq is:

- reflexive: $\forall A \in \mathcal{P}(Y) \ ARA \iff A \subseteq A$;
- anti-symmetric: $\forall A, B \in \mathcal{P}(Y) \ ARB \land BRA \iff A \supseteq B \land B \supseteq$ $A \implies A = B$;
- transitive: $\forall A, B, C \in \mathcal{P}(Y)$ $ARB \land BRC \iff A \supseteq B \land B \supseteq C \implies$ $A \supseteq C$.

All the above examples are also known as partially ordered sets.

- ¹ We can look at this definition as $R \subseteq X \times X$. Under such a definition, we would have
- (Reflexive) $\forall x \in X \ (x, x) \in R$;
- (Symmetric) $\forall x, y \in X (x, y) \in$
- (Anti-symmetric) $\forall x, y \in$
- (Transitive) $\forall x, y, z \in$

Definition 14 (Partially Ordered Sets)

The set X with the relation R on X is called a partially ordered set (or a poset) if R is

- reflexive;
- anti-symmetric; and
- transitive.

We denote a poset by (X, R).

The "partial" in 'partially ordered" indicates that not every pair of elements need to be comparable, i.e. there may be pairs for which neither precedes the other (anti-symmetry).

66 Note

If (X, R) *is a poset, then if* $A \subseteq X$ *, and* $R_1 = R \upharpoonright_{A \times A}$ *, then* (A, R_1) *is also a poset.*

Example 3.2.4

How many possible relations can we define on these sets to make them into posets?

1.
$$X = \emptyset$$

Solution

We have that $R = \emptyset \times \emptyset$, and so the only relation we have is an empty relation. Then it is vacuously true that (X, R) a poset.

2.
$$X = \{x\}$$

Solution

We have that $R = X \times X = \{(x, x)\}$. It it clear that (X, R) is a poset.

3.
$$X = \{x, y\}$$

Solution

There are 3 possible relations:

- a relation where *xRx* and *yRy*;
- a relation where *xRy*; or
- a relation where yRx.

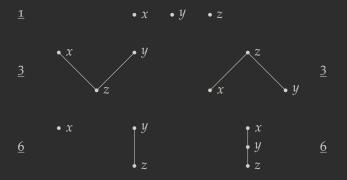
3 possibilities represented as graphs (known as Hasse diagram), separated by lines:



4.
$$X = \{x, y, z\}$$

Solution

The following are all the possibilities represented by graphs, where the underlined numbers represent the number of ways we can rearrange the elements for unique relations:



Therefore, we see that there are a total of

$$1+3+3+6+6=19$$
 relations.

Exercise 3.2.1

How many possible relations can we define on a set of 6 elements to the set into a poset?

Solution

to be added

Definition 15 (Totally Ordered Sets / Chains)

The set X with the relation R on X is called a **totally ordered set** (or a chain) if (X, R) is a poset with the exception that, for any $x, y \in X$, either xRy or yRx but not both.

Definition 16 (Bounds)

Let (X, \leq) be a poset. Let $A \subset X$. We say $x_0 \in X$ is an upper bound for A if

$$\forall a \in A \quad a \leq x_0.$$

If A has an upper bound, we say that A is bounded above. If A is bounded above, then x_0 is the least upper bound (or supremum) of A is for any $x_1 \in X$ that is an upper bound of A, we have

$$x_0 \leq x_1$$
.

We write $x_0 = \text{lub}(A) = \sup(A)$. If $\sup(A) \in A$, then $\sup(A) = \max(A)$ is the maximum of A.

We can analogously define for:

upper bound \rightarrow lower bound

bounded above \rightarrow bounded below

least upper bound, lub \rightarrow greatest lower bound, glb

supremum, $sup \rightarrow infimum$, inf

 $maximum, max \rightarrow minimum, min$

66 Note

By anti-symmetry of posets, we have that max, sup, min, inf are all unique if they exists.

Example 3.2.5 (Least Upper Bound Property of R)

Let $X = \mathbb{R}$, and \leq be the order that we have defined. Every bounded non-empty subset of X has a supremum.

Example 3.2.6

Let $Y \neq \emptyset$, and $X = \mathcal{P}(Y)$, and \subseteq the ordering by inclusion. We know that Y is the maximum element of (X, \subseteq) . Then the collection $\{A_{\alpha}\}_{{\alpha} \in I} \subset \mathcal{P}(Y)$ is bounded above by Y, and we have that

$$\sup (\{A_{\alpha}\}_{\alpha \in I}) = \bigcup_{\alpha \in I} A_{\alpha}$$

$$\inf (\{A_{\alpha}\}_{\alpha \in I}) = \bigcap_{\alpha \in I} A_{\alpha}$$

Now if $Y = \emptyset$, we would end up having

$$\sup (\{A_{\alpha}\}_{\alpha \in I}) = \emptyset$$

$$\inf (\{A_{\alpha}\}_{\alpha \in I}) = X$$

But this makes sense, since the empty set would be the least of upper bounds, and since $X = \mathcal{P}(Y)$ would have to be the greatest of lower bounds.

4 Lecture 4 Sep 14th

4.1 Zorn's Lemma

Definition 17 (Maximal Element)

Let (X, \leq) be a poset. An element $x \in X$ is maximal if whenever $y \in X$ is such that $x \leq y$, we must have y = x.

Example 4.1.1

Looking back at Example 3.2.4, on the set $X = \{x, y, z\}$, we have that the maximal element in each possible poset is/are:

This shows to us that the maximal element does not have to be unique.



Example 4.1.2

- Given $X \neq \emptyset$, the maximal element of the poset $(\mathcal{P}(X), \subseteq)$ is X.
- Given $X \neq \emptyset$, the maximal element of the poset $(\mathcal{P}(X), \supseteq)$ is \emptyset .
- The poset (\mathbb{R}, \leq) has no maximal element.

▼ Axiom 4 (Zorn's Lemma)

If (X, \leq) is a non-empty poset such that every chain $S \subset X$ has an upper bound, then (X, \leq) has a maximal element.

■ Theorem 5 (★ Non-Zero Vector Spaces has a Basis)

Every non-zero vector space, V, has a basis.

Proof (★)

Let

 $\mathcal{L} := \{ A \subset V \mid A \text{ is linearly independent } \}.$

Note that $\mathcal{L} \neq \emptyset$ since $V \neq \{0\}$. Now order elements of \mathcal{L} with \subseteq . It suffices to show that (\mathcal{L}, \subseteq) has a maximal element, since this maximal element must be a basis. Otherwise, we would contradict the maximality of such an element.¹

Now let $S = \{A_{\alpha}\}_{{\alpha} \in I}$ be a chain in \mathcal{L} . Let

$$A_0 = \bigcup_{\alpha \in I} A_{\alpha}.$$

Require clarification before proceeding...

The flow of this proof is a typical approach when Zorn's Lemma is involved.

¹ This is the key to this proof.

Definition 18 (Well-Ordered)

We say that a poset (X, \leq) is well-ordered if every non-empty subset $A \subset X$ has a least/minimal element in A.

Exercise 4.1.1

Prove that well-ordered sets are chains.

Example 4.1.3

 (\mathbb{N}, \leq) is well-ordered.

▼ Axiom 6 (Well-Ordering Principle)

Every non-empty set can be well-ordered.

Theorem 7 (Axioms of Choice and Its Equivalents)

TFAE:

- 1. Axiom of Choice, **Ū** Axiom 2
- 2. Zorn's Lemma, **U** Axiom 4
- 3. Well-Ordering Principle, \mathbf{V} Axiom 6.

Exercise 4.1.2 Prove 👤 Theorem 7

Proof

 $(3) \implies (1)$ is simple; let the choice function be such that we pick the minimal element from each set among a non-empty collection of non-empty sets. It is clear that the product of these sets will always have an element, in particular the tuple where each component is the minimal element of each set.

The rest will be added once I've worked it out

Example 4.1.4

Let $X = \mathbb{Q}$. Let $\phi : \mathbb{Q} \to \mathbb{N}$ be defined such that

$$\phi\left(\frac{m}{n}\right) = \begin{cases} 2^m 5^n & m > 0\\ 1 & m = 1\\ 3^{-m} 7^n & m < 0 \end{cases}$$

By the unique prime factorization of natural numbers (or Fundamental Theorem of Arithmetic), we have that ϕ is injective. In fact,

$$r \leq s \iff \phi(r) \leq \phi(s)$$
,

showing to us that we have a well-ordering on Q.

4.2 Cardinality

4.2.1 Equivalence Relation

Definition 19 (Equivalence Relation)

Let X be non-empty set. A relation \sim on X is an equivalence relation if it is

- reflexive;
- symmetric; and
- transitive.

Definition 20 (Equivalence Class)

Let X be a non-empty set, and $x \in X$. An equivalence class of x under the equivalence relation \sim is defined as

$$[x] := \{ y \in X \mid x \sim y \}.$$

66 Note

Note that we either have [x] = [y] or $[x] \cap [y] = \emptyset$. This is sensible, since if $w \in [x]$, then $w \sim x$. If $w \in [y]$, then we are done. If $w \notin [y]$, suppose $\exists v \in [y]$ such that $w \sim v$, which then implies $w \in [y]$ which is a contradiction.

This results shows to us that

$$X = \bigcup_{x \in X} [x],$$

or in words, equivalence classes partition the set.

Definition 21 (Partition)

Let $X \neq \emptyset$ *. A partition of* X *is a collection* $\{A_{\alpha}\}_{\alpha \in I} \subset \mathcal{P}(X)$ *such that*

- 1. $A_{\alpha} \neq \emptyset$;
- 2. $A_{\alpha} \cap A_{\beta} = \emptyset$ if $\alpha \neq \beta$ in I; and
- 3. $X = \bigcup_{\alpha \in I} A_{\alpha}$.

With this, we have ourselves another method to show that \sim is an equivalence relation.

• Proposition 8 (Characterization of An Equivalence Relation)

If $\{A_{\alpha}\}_{{\alpha}\in I}$ is a partition of X and $x\sim y\iff x,y\in A_{\alpha}$, then \sim is an equivalence relation.

The proof of this statement has been stated above.

Similar to when we defined partial orders, we can ask ourselves the following question:

Example 4.2.1

How many equivalence relations are there on the set $X = \{1, 2, 3\}$?

Solution

Note that we can partition *X* as

$$\{\{1\},\{2\},\{3\}\},\{\{1,2,3\}\},$$

and

which we can rearrange in 3 different ways. Therefore, there are 5 different equivalence relations that we can define on *X*.

Example 4.2.2

Let *X* be any set. Consider $\mathcal{P}(X)$. Define \sim on $\mathcal{P}(X)$ by

$$A \sim B \iff \exists f : A \to B$$

such that f is surjective³. It is easy to verify that \sim is an equivalence relation.

² By **6** Proposition 8, this question is equivalent to asking for the number of partitions we can create from the set X. The study of counting partitions is what is covered by Bell's Number.

 $^{^{3} \}sim$ partitions *X* into sets that have the same number of elements.

5 Lecture 5 Sep 17th

5.1 Cardinality (Continued)

Definition 22 (Finite Sets)

A set X is finite if $X = \emptyset$ or if $X \sim \{1, 2, ..., n\}$ for some $n \in \mathbb{N}$, where \sim is the equivalence relation defined in Example 4.2.2.

Definition 23 (Cardinality)

If $X \sim n$, we say X has cardinality n and write |X| = n. We also let $|\emptyset| = 0$.

Now a good question here is: if $n \neq m$, is $\{1, 2, ..., n\} \sim \{1, 2, ..., m\}$?

Theorem 9 (Pigeonhole Principle)

The set $\{1, 2, ..., n\}$ is not equivalent to any of its proper subset.

Proof

We shall prove this by induction on n.

Base case: $\{1\} \sim \emptyset$.

Assume that the statement holds for $\{1,...,k\}$. Suppose we have an

This is a **proof by contradiction**, using the fact that we cannot find an injective function from a "larger" set to a "smaller" set.

We can assume that the function f is not surjective, since if the larger set is indeed equivalent to the smaller set, then it should not matter if f is surjective or not. In particular, we only require that there be an injective function

Requires clarification and confirmation of proof.

injective function

$$f: \{1, 2, ..., k, k+1\} \rightarrow \{1, 2, ..., k, k+1\}$$

that is not surjective.

Case 1: $k + 1 \notin \text{Range}(f)$, where Range(f) is the range of f. Then we have

Note: ↑ is the restriction sign.

$$f \upharpoonright_{\{1,...,k\}} : \{1,...,k\} \to \{1,...,k\} \setminus \{f(k+1)\}.$$

But *f* is an injective function and clearly

$$\{1,...,k\} \setminus \{f(k+1)\} \subseteq \{1,...,k\},$$

a contradition.

Case 2: $k + 1 \in \text{Range}(f)$. Then $\exists j_0 \in \{1, ..., k, k + 1\}$ such that $f(j_0) = k + 1$, and since f is not surjective, $\exists m \in \{1, ..., k\}$ such that $m \notin \text{Range}(f)$. Then consider a new function $g : \{1, ..., k, k + 1\} \rightarrow \{1, ..., k\}$ such that

$$g(a) = \begin{cases} m & a = k+1 \\ f(k+1) & a = j_0 \\ f(a) & a \neq j_0, k+1 \end{cases}$$

► Corollary 10 (Pigeonhole Principle (Finite Case))

If the set X is finite, then X is not equivalent to any proper subset.

Exercise 5.1.1

Prove **>** *Corollary* 10.

Sketch of proof: $\{1,...,n\} \longrightarrow \{1,...,n\}$ $f^{-1} \downarrow f \qquad \qquad \uparrow f$ $Y = \frac{1-1}{2} A \subset Y$

Definition 24 (Infinite Sets)

X is *infinite* if it is not finite.

Example 5.1.1

¹ **U** Axiom 3 ahoy!

Observe that we can construct a function $f: N \to \{2,3,...\}$ by f(n) =n+1. It is clear that f is a bijective function, and so $\mathbb{N} \sim \{2,3,...\}$.

• Proposition 11 (N is the Smallest Infinite Set)

Every infinite set contains a subset $A \sim \mathbb{N}$.

Proof

Suppose X is infinite. Let

$$f: \mathcal{P}(X) \setminus \{\emptyset\} \to X$$

such that for $S \subset X$ where $S \neq \emptyset$, $f(S) \in S$ 1. Let $x_1 = f(X)$. Let $x_2 = f(X \setminus \{x_1\})$. Recursively, define

$$x_n = f(X \setminus \{x_1, ..., x_{n-1}\}).$$

This gives us a sequence

$$X \supset S = \{x_1, ..., x_n, ...\}$$

which is equivalent to \mathbb{N} via the map $n \mapsto x_n$.

Corollary 12 (Infinite Sets are Equivalent to Its Proper Subsets)

Every infinite set X is equivalent to a proper subset of X.

Proof

Given such an X, we construct a sequence $\{x_n\}$ as in the previous proof. Define $f: X \to X \setminus \{x_n\}$ by

$$f(x) = \begin{cases} x_{n+1} & x \in \{x_n\} \\ x & x \notin \{x_n\}. \end{cases}$$

Clearly so, f is injective.

Definition 25 (Countable)

We say that a set is **countable** (or **denumerable**) is either X is finite or if $X \sim \mathbb{N}$. If $X \sim \mathbb{N}$, we can say that X is **countably infinite** and write $|X| = |\mathbb{N}| = \aleph_0$.

Definition 26 (Smaller Cardinality)

Given 2 sets X, Y, we write

$$|X| \leq |Y|$$

if $\exists f: X \rightarrow Y$ *injective.*

• Proposition 13 (Injectivity is Surjectivity Reversed)

TFAE:

- 1. $\exists f: X \rightarrow Y \text{ injective}$
- 2. $\exists g: Y \to X \ \overline{surjective}$

Proof

 $(1) \implies (2)$: Define

$$g(y) = \begin{cases} x & \exists x \in X \ f(x) = y \\ x_0 & \text{any } x_0 \in X \end{cases}$$

Clearly *g* is surjective.

(2) \implies (1): Since *g* is surjective, for each $x \in X$, we have that²

$$g^{-1}(|x|) = \{ y \in \Upsilon : g(y) = x \} \neq \emptyset.$$

By the Axiom of Choice, there exists a choice function $h: \mathcal{P}(Y) \setminus \{\emptyset\} \to Y$ such that for each $A \subset Y$, $h(A) \in A$. Then, let $f: X \to Y$ such that

$$f(x) = h(g^{-1}(\{x\})).$$

² The idea here is to collect the preimages into a set, and use the choice function as an injective map.

Clearly so, f is injective.

66 Note

Note that we have $|\mathbb{N}| \leq |\mathbb{Q}|$, since we can define an injective function $f: \mathbb{N} \to \mathbb{Q}$ such that $f(n) = \frac{n}{1}$.

We have also shown that $|Q| \leq |\mathbb{N}|$ using our injective function $g: \mathbb{Q} \to \mathbb{N}$, given by

$$g\left(\frac{m}{n}\right) = \begin{cases} 2^{m}3^{n} & m > 0\\ 1 & m = 0\\ 5^{-m}7^{n} & m < 0 \end{cases}$$

Question: Is $|\mathbb{N} = |\mathbb{Q}||$? In other words, given $|X| \leq |Y| \wedge |Y| \leq |X|$,

6 Lecture 6 Sep 19th

6.1 Cardinality (Continued 2)

Before delving into resolving our last question in the previous lecture, note the following:

66 Note

Suppose $f: X \to Y$ is bijective. Let $A \subseteq B$, then

$$f(B \setminus A) = f(B) \setminus f(A)$$
.

Prove this observation as an exercise:

Exercise 6.1.1 Prove the note on the left.

■ Theorem 14 (★★★ Cantor-Schröder-Bernstein Theorem (CSB))

Let $A_2 \subset A_1 \subset A_0 = A$. Assume that $A_2 \sim A_0$. Then $A_0 \sim A_1$.

Proof

Let $\phi: A_0 \to A_2$ be bijective, by assumption. Since $A_1 \subset A_1$, let $A_3 = \phi(A_1) \subset A_2$, and since $A_2 \subset A_0$, let $A_4 = \phi(A_2) \subset A_3$. Recursively so, let

$$A_{n+2} = \phi(A_n)$$

Notice that

$$A_0 = (A_0 \setminus A_1) \cup (A_1 \setminus A_2) \cup (A_2 \setminus A_3) \cup (A_3 \setminus A_4) \cup \dots \bigcap_{n=0}^{\infty} A_n$$
$$A_1 = (A_1 \setminus A_2) \cup (A_2 \setminus A_3) \cup (A_3 \setminus A_4) \cup (A_4 \setminus A_5) \cup \dots \bigcap_{n=1}^{\infty} A_n$$

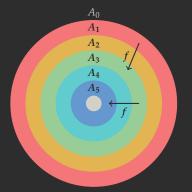


Figure 6.1: The core idea of the proof for Cantor-Schröder-Bernstein Theorem

Observe that

$$\bigcap_{n=0}^{\infty} A_n = \bigcap_{n=1}^{\infty} A_n.$$

¹Define $f: A \to A_1$ by

$$f(x) = \begin{cases} x & x \in \bigcap_{n=0}^{\infty} A_n \\ x & x \in A_{2k+1} \setminus A_{2k+2}, \ k = 0, 1, 2, \dots \\ \phi(x) & x \in A_{2k} \setminus A_{2k+1}, \ k = 0, 1, 2, \dots \end{cases}$$

¹ Here, we employ the idea from Figure 6.1.

► Corollary 15 (Cantor-Schröder-Bernstein Theorem - Restated)

If $A_1 \subset A \wedge B_1 \subset B \wedge A \sim B_1 \wedge B \sim A_1$, then $A \sim B$.²

² This is equivalent to the statement $|A| < |B| \land |B| < |A| \implies |A| = |B|$

Proof

By assumption, let $f:A\to B_1$ be bijective, and let $g:B\to A_1$ be bijective. Let $A_2=g(B_1)\subseteq A_1\subset A$ Let $A_2=g(B_1)\subseteq A_1\subset A$. Then the composite function $g\circ f:A\to A_2$ is bijective, and so $A\sim A_2$. By \blacksquare Theorem 14, we have

$$A \sim A_2 \sim A_1 \sim B$$
.

Example 6.1.1

Our question from last lecture now has an answer: by \blacksquare Theorem 14, we have that $|\mathbb{Q}| = |\mathbb{N}|.^3$

³ Now that we know that they have the ssme cardinality:

• Proposition 16 (Denumerability Check)

If X is infinite, *then*

$$|X| = |\mathbb{N}| = \aleph_0 \iff \exists f : X \to \mathbb{N} \text{ bijective.}$$

Exercise 6.1.2

Find a bijection between \mathbb{Q} and \mathbb{N} .

Proof

(
$$\Longrightarrow$$
) is immediate. For (\Longleftrightarrow), suppose $f:X\to\mathbb{N}$, which implies that $|X|\leq |\mathbb{N}|$. By \bullet Proposition 11, $|\mathbb{N}|\leq |X|$. Therefore, $|X|=|bbN|=\aleph_0$.

Example 6.1.2

 $\mathbb{N} \times \mathbb{N}$ is countable. The function

$$f: \mathbb{N} \times \mathbb{N} \times \mathbb{N}$$
 given by $f(m,n) = 2^n 3^m$

is injective.

Definition 27 (Uncountable)

A set X is **uncountable** if it is not countable.

■ Theorem 17 (Cantor's Diagonal Argument)

(0,1) is uncountable.

Proof

Suppose, for contradiction, that (0,1) is countable. Then we can write

$$a_1 = .a_{11}a_{12}a_{13}...$$

 $a_2 = .a_{21}a_{22}a_{23}...$
 \vdots
 $a_n = .a_{n1}a_{n2}a_{n3}...$

in (0,1). This representation is unique if we do not allow the representation to end in a string of 9's. Let $b \in (0,1)$, expressed as $b = .b_1b_2b_3...$ such that

$$b_i = \begin{cases} 5 & a_i \neq 5 \\ 2 & a_i = 5 \end{cases}$$

Then $b \notin (0,1)$, a contradiction⁴.

⁴ Require more explanation.

► Corollary 18 (Uncountability of ℝ)

 \mathbb{R} is uncountable.

Proof

Let $f:(0,1)\to\mathbb{R}$ be given by

$$f(x) = \tan\left(\pi x - \frac{\pi}{2}\right).$$

Clearly so, (0,1) is bijective.

66 Note

We denote $|\mathbb{R}| = c$.

QUESTION: Given sets X, Y, is it always true that either⁵

- 1. |X| = |Y|;
- 2. |X| < |Y|: or
- 3. |Y| < |X|.

 5 As compare to \leq , < implies that there is no surjection from the set on the LHS to the RHS.

7 Lecture 7 Sep 21st

7.1 Cardinality (Continued 3)

■ Theorem 19 (★ Comparability of Cardinals)

If X and Y are non-empty, then either

$$|X| \le |Y| \lor |Y| \le |X|.$$

Proof

Let

$$S = \{(A, B, f) \mid A \subseteq X, B \subseteq Y, f : A \rightarrow B \text{ bijective } \}.$$

Note that $S \neq \emptyset$, since X and Y are non-empty, and so we can have f(a) = b for $A = \{a\} \subset X$ and $B = |b| \subset Y$. We order S as follows: we say

$$(A_1, B_1, f_1) \le (A_2, B_2, f_2)$$

if

$$A_1 \subseteq A_2$$
, $B_1 \subseteq B_2$, $f_1 = f_2 \upharpoonright_{A_1}$.

Let $C = \{(A_{\alpha}, B_{\alpha}, f_{\alpha})\}_{\alpha \in I}$ be a chain in (S, \leq) . Let $A_0 = \bigcup_{\alpha \in I} A_{\alpha}$, $B_0 = \bigcup_{\alpha \in I} B_{\alpha}$, and define $f_0 : A_0 \to B_0$ by

$$f_0(x) = f_{\alpha_0}(x)$$
 if $x \in A_{\alpha_0}$.

Now if $x \in A_{\alpha_1}$, $x \in A_{\alpha_2}$ and

$$(A_{\alpha_1}, B_{\alpha_1}, f_{\alpha_1}) \leq (A_{\alpha_2}, B_{\alpha_2}, f_{\alpha_2}),$$

¹ We want to use the maximal element to obtain our result. To that end, we need Zorn's Lemma. So we need *S* to build this up.

we have that

$$f_{\alpha_1}(x) = f_{\alpha_2} \upharpoonright_{A_{\alpha_1}} (x) = f_{\alpha_2}(x),$$

i.e. f_0 is well-defined.

Claim 1: $f_0: A_0 \rightarrow B_0$ is injective.

Let $x_1, x_2 \in A_0$ such that $x_1 \neq x_2$.

$$\implies \exists \alpha_1, \alpha_2 \in I \ x_1 \in A_{\alpha_1} \land x_2 \in A_{\alpha_2} \land A_{\alpha_1} \subseteq A_{\alpha_2} \text{ (wlog)}$$

$$\implies x_1.x_2 \in A_{\alpha_2}$$

$$\implies (:: f_{\alpha_2} \text{ injective } \implies f_{\alpha_2}(x_1) \neq f_{\alpha_2}(x))$$

$$\implies f_0(x_1) \neq f_0(x_2) \implies f_0$$
 injective.

Claim 2: $f_0: A_0 \to B_0$ is surjective.

Let $y_0 \in B_0$

$$\implies \exists \alpha_0 \in I \ y_0 \in B_{\alpha_0}$$

$$\implies \exists x_0 \in A_{\alpha_0} \ f_{\alpha_0}(x_0) = y_0 \ (\because f_{\alpha_0} \ \text{surjective})$$

$$\implies f_0(x_0) = y_0$$

 \therefore (A_0, B_0, f_0) is an upper bound for C. Then by Zorn's Lemma, (S, \leq) has a maximal element (A, B, f).

Case 1: If A = X, then injectivity of f implies $|X| \leq |Y|$.

Case 2: If B = Y, then surjectivity of f implies $|Y| \le |A| \le |X|$.

Case 3: If $A \neq X \land B \neq Y$, then $X \setminus A \neq \emptyset \land Y \setminus B \neq \emptyset$. Let $x_0 \in X \setminus A$, $y_0 \in Y \setminus B$. Let $A^* = A \cup \{x_0\}$, $B^* = B \cup \{y_0\}$, and $f^* : A^* \to B^*$ such that

$$f^*(x) = \begin{cases} f(x) & x \in A \\ y_0 & x = x_0 \end{cases}$$

Then $(A, B, f) \le (A^*, B^*, f^*)$, contradicting maximality.

7.1.1 Cardinal Arithmetic

Sum of Cardinals Observe that if $X = \{x_1, ..., x_n\}$, $Y = \{y_1, ..., y_m\}$, and $X \cap Y = \emptyset$, then |X| = n, |Y| = m and $|X \cup Y| = n + m$. This motivates us to provide the following definition:

Assume that X and Y are such that X \cap *Y* = \emptyset *. We define*

$$|X| + |Y| = |X \cup Y|.$$

Question: So what about $\aleph_0 + \aleph_0$?

A thought that motivates us to give the following answer lies in the observation that: if *X* is the set of even natural numbers and *Y* the odd natural numbers, then $X \cup Y$ is the set of all natural numbers. All three sets are countably infinite, i.e. they have cardinality \aleph_0 .

QUESTION: What about c + c?

A similar motivation comes from the observation that: given X =(0,1) and Y = (1,2), we have

$$c = |X| \le |X| + |Y| \le |R| = c$$

and so
$$|X| = |Y| = c \implies |X \cup Y| = c$$
.

Theorem 20 (Sums of Cardinals)

Given sets X and Y, if X is infinite, then

1.
$$|X| + |X| = |X|$$

2.
$$|X| + |Y| = \max(|X|, |Y|)$$

Multiplication of Cardinals Given

$$X = \{x_1, x_2, ..., x_n\}$$

 $Y = \{y_1, y_2, ..., y_m\}$

we have that

$$X \times Y = \{(x_i, y_j) \mid i = 1, 2, ..., n, j = 1, 2, ..., m\}$$

and so

$$|X \times Y| = nm$$
.

Exercise 7.1.1

Prove Prove Theorem 20 as an exercise.

Definition 29 (Multiplication of Cardinals)

Given sets X and Y, define

$$|X||Y| = |X \times Y|.$$

Example 7.1.1

We have $|\mathbb{N} \times \mathbb{N}| = \aleph_0$ since the function $f : \mathbb{N} \times \mathbb{N} \to \mathbb{N}$ given by

$$f(n,m) = 2^n 3^m$$

is injective.

Question: What about $c \cdot c$?

■ Theorem 21 (Multiplication of Cardinals)

If X *is infinite and* $Y \neq \emptyset$ *, then*

- $|X \times X| = |X| \implies |X| |X| = |X|$;
- $|X \times Y| = \max(|X|, |Y|)$.

Exercise 7.1.2

Prove Prove Theorem 21 as an exercise.

8 Lecture 8 Sep 24th

8.1 Cardinality (Continued 4,

8.1.1 Cardinal Arithmetic (Continued)

Exponentiation of Cardinals Recall if $\{Y_x\}_{x\in X}$ is a collection of nonempty sets, we have¹

$$\prod_{x \in X} Y_x = \{ f : X \to \bigcup_{x \in X} Y_x \mid f(x) \in Y_x \}.$$

Now if $Y = Y_x$ for all $x \in X$, we have

$$Y^X = \prod_{x \in X} = \{f : X \to Y\}.$$

Example 8.1.1

Given

$$Y = \{1, ..., m\}$$
 $X = \{1, ..., n\}$

we have

$$Y^X = \{f : \{1, ..., n\} \rightarrow \{1, ..., m\}\}.$$

Observe that Y^X is similar to Y^n . So $|Y^X| = m^n$.

¹ This should remind you of **♥** Axiom 3

² Need better explanation.

Definition 30 (Exponentiation of Cardinals)

Given sets X and Y, define

$$\left|Y\right|^{\left|X\right|}:=\left|Y^{X}\right|.$$

■ Theorem 22 (Exponentiation of Cardinals)

If X, Y, \overline{Z} are non-empty sets, then

•
$$|Y|^{|X|} \cdot |Y|^{|Z|} = |Y|^{|X|+|Z|}$$
;

$$\bullet \ \left(|Y|^{|X|} \right)^{|Z|} = |Y|^{|X| \cdot |Z|}$$

\blacksquare Theorem 23 (2^{\aleph_0} = c)

We have that $2^{\aleph_0} = c$.

Exercise 8.1.1

Prove Prove Prove 22.

Proof

Note that $2^{\aleph_0} = |\{0,1\}^{\mathbb{N}}|$, where³

$$|\{0,1\}^{\mathbb{N}}| = |\{f : \mathbb{N} \to \{0,1\}| = |\{\{a_n\}_{n=1}^{\infty} \mid a_i = 0,1\}|$$

Given a sequence $\{a_n\} \in \{0,1\}^{\mathbb{N}}$, define $\phi : \{0,1\}^{\mathbb{N}} \to (0,1)$ such that⁴

$$\phi\left(\left\{a_n\right\}\right) := \sum_{i=1}^{\infty} \frac{a_n}{3^n}.$$

which is injective since there are no trailing 2's. Therefore $2^{\aleph_0} \le c$.

Given $\alpha \in (0,1)$, let⁵

$$\alpha = \sum_{n=1}^{\infty} \frac{b_n}{2^n},$$

where $b_n = 0, 1$. Let $\psi : (0, 1) \to \{0, 1\}^{\mathbb{N}}$ such that

$$\psi(\alpha) = \psi\left(\sum_{i=1}^{\infty} \frac{b_n}{2^n}\right) = \{b_n\}$$

Then ψ is injective, and so $c \leq 2^{\aleph_0}$. Thus $2^{\aleph_0} = c$ as required.

This requires closer studying.

³ Explain 2nd equality.

⁴ This is a base 3 representation (of what?)

⁵ This is a base 2 representation.

Example 8.1.2

Find $\left|\aleph_0^{\aleph_0}\right|$ and c^{\aleph_0} .

Solution

We have that

$$c = 2^{\aleph_0} \le \aleph_0^{\aleph_0} \le c^{\aleph_0} = \left(2^{\aleph_0}\right)^{\aleph_0} = 2^{\aleph_0 \cdot \aleph_0} = 2^{\aleph_0} = c$$

Example 8.1.3

Show $|\mathcal{P}(X)| = 2^{|X|} = |2^X|$.

Solution

Given $f: X \to \{0,1\}$, let⁶

$$A = \{x \in X \mid f(x) = 1\} \subset X.$$

Define $\Gamma: 2^X \to \mathcal{P}(X)$ by

$$\Gamma(f) = f^{-1}(|1|)$$

 Γ is injective⁷.

Conversely, given $A \subset X$, define the characteristic function

$$\chi_A(x) = egin{cases} 1 & x \in A \ 0 & x
otin A \end{cases} \in 2^{\mathrm{X}}.$$

Then define $\Phi: P(A) \to 2^X$ such that

$$\Phi(A) = \chi_A.$$

Clearly so, Φ is injective.

Theorem 24 (Russell's Paradox)

For any X, we have $|X| < |\mathcal{P}(X)| = 2^{|X|}$.

Proof

Let $f: X \to \mathcal{P}(X)$ be $f(X) = \{x\}$. Clearly, f is injective, and so $|X| < |\mathcal{P}(X)|$.

Claim: $\nexists g: X \to \mathcal{P}(X)$ surjective.

Suppose not. Let⁸

$$A = \{ x \in X \mid x \notin g(x) \}$$

Pick $x_0 \in X$ with $g(x_0) = A$. Now if $x_0 \in A$, then $x_0 \in g(x_0)$, but this implies that $x \notin A$, a contradiction.

So $x_0 \notin A$, i.e. $x \notin g(x_0)$, which in turn implies that $x \in A$, yet

⁶ A is a collection of all x's that gets mapped to f.

7 Why?

⁸ By the **Bounded Separation Axiom** (see ZF Set Theory), this is a set, and since it is a subset of X, it is a valid element of $\mathcal{P}(X)$. Thus, we can consider such a set.

48 Lecture 8 Sep 24th - Cardinality (Continued 4)

another contradiction. Therefore such a function g cannot exist, as claimed.

Therefore, we have $|X| < |\mathcal{P}(X)|$ as required.

QUESTION: Is there anything between \aleph_0 and c?

■ Axiom 25 (Continuum Hypothesis)

If $\aleph_0 \leq \gamma \leq c$, then either $\gamma = \aleph_0$ or $\gamma = c$.

■ Axiom 26 (Generalized Continuum Hypothesis)

If
$$|X| \le \gamma \le 2^{|X|}$$
, then either $\gamma = |X|$ or $\gamma = 2^{|X|}$.

In this course, we shall assume that the Continuum Hypothesis is true.

9 Lecture 9 Sep 26th

9.1 Introduction to Metric Spaces

■ Definition 31 (Metric & Metric Space)

Given a set X, a *metric* on X is a function $d: X \times X \to \mathbb{R}$ such that

- 1. (positive definiteness) $d(x,y) \ge 0$ and $d(x,y) = 0 \iff x = y$;
- 2. (symmetry) d(x,y) = d(y,x); and
- 3. (triangle inequality) $d(x,y) \le d(x,z) + d(z,y)$.

The pair (X, d) is called a metric space.

Example 9.1.1 (Standard Metric on \mathbb{R})

Let $X = \mathbb{R}$, and let d(x, y) = |x - y|.

Clearly so, the first 2 criterias are satisfied:

- $|\cdot| \ge 0$ and $|x-y| = 0 \iff x = y$; and
- |x y| = |y x|

The triangle inequality property is the usual triangle inequality of the absolute value function, i.e.

$$|x-y| \le |x| + |y|.$$

QUESTION: For an arbitrary set X, can we define a metric on X? The following example shows that we can,

Example 9.1.2 (Discrete Metric)

Remark

A metric is an abstract notion of distance.

Let *X* be any set. We can simply define

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

This metric clearly satisfies all 3 criterias of being a metric:

- $d: X \times X \to \{0,1\}$ and so $d(x,y) \ge 0$, and by definition, we have $d(x,y) = 0 \iff x = y$;
- By definition, d(x,y) = d(y,x) as it does not matter how the pair is ordered; and
- Since $d(x,y) \ge 0$, we have that $d(x,y) \le d(x,z) + d(y,z)$.

Example 9.1.3 (Euclidean Metric / 2-metric on \mathbb{R}^n)

Let $X = \mathbb{R}^n$. Let $\vec{x} = \{x_1, x_2, ..., x_n\}$ and $\vec{y} = \{y_1, y_2, ..., y_n\}$. Define

$$d_2(\vec{x}, \vec{y}) = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$$

Note that in \mathbb{R}^2 , this is our regular (Euclidean) distance between two points.

It is not difficult to see that d_2 satisfies the first 2 criterion to being a metric:

- d_2 is the square root of the sum of squares, and so $d_2(\vec{x}, \vec{y}) \ge 0$ for any $\vec{x}, \vec{y} \in \mathbb{R}^n$, and $d_2(\vec{x}, \vec{y}) = 0 \iff \forall i \in \{1, ..., n\} \ x_i = y_i \iff \vec{x} = \vec{y}$:
- Since $(x_i y_i)^2 = (y_i x_i)^2$ for any $x_i, y_i \in \mathbb{R}$, we have that $d_2(\vec{x}, \vec{y}) = d_2(\vec{y}, \vec{x})$.

However, it is not immediately clear that d_2 satisfies the triangle inequality criterion, especially if $n \geq 3$. If n = 2, heuristically, the triangle inequality simply tells that the length of any one side of a triangle is less than or equal to the sum of the other two, e.g. Figure 9.1.

$\vec{x} + \vec{y}$ \vec{x}

Figure 9.1: A visualization of the triangle inequality in \mathbb{R}^2 .

Remark

Many of the important examples of metric spaces are vector spaces with an abstract length function, or **norm**.

Definition 32 (Norm & Normed Linear Space)

Given a vector space V (usually over \mathbb{R}), a **norm** on V is a function

$$\|\cdot\|:V\to\mathbb{R}$$

such that

- 1. (positive definiteness) $||v|| \ge 0$ and $||v|| = 0 \iff v = 0$;
- 2. (scalar multiplication) $\|\alpha \cdot v\| = |\alpha| \|v\|$; and
- 3. (triangle inequality) $||v + w|| \le ||v|| + ||w||$.

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

Remark

Given a normed linear space $(V,\|\cdot\|)$, a natural metric, $d_{\|\cdot\|}$, on V induced by $\|\cdot\|$ can be defined as

$$d_{\|\cdot\|}(x,y) = \|x-y\|.$$

Exercise 9.1.1

Prove that $d_{\|.\|}$ *is indeed a metric.*

Proof (Exercise 9.1.1)

- 1. (positive definiteness) It is clear from the definition of a norm that $d_{\|\cdot\|}(x,y) = \|x-y\| \ge 0$, and $d_{\|\cdot\|}(x,y) = 0 \iff x-y = 0$
- 2. (symmetry) Symmetry follows simply from definition, as ||x y|| =
- 3. (triangle inequality) For $x, y, z \in V$, we have

$$d_{\|\cdot\|}(x,y) = \|x-y\| = \|x-z+z-y\|$$

 $\leq \|x-z\| + \|z-y\|$: triangle inequality of norms
 $= \|x-z\| + \|y-z\|$: symmetry
 $= d_{\|\cdot\|}(x,z) + d_{\|\cdot\|}(y,z)$

Example 9.1.4 (Euclidean Norm)

Let $X = \mathbb{R}^n$, and $\vec{x} = (x_1, ..., x_n) \in \mathbb{R}^2$. Define $\|\cdot\|_2$ such that

$$\|(x_1,...,x_n)\|_2 = \sqrt{\sum_{i=1}^n x_i^2}$$

From Example 9.1.3, we are given the triangle inequality property, in

which we have yet to prove. Positive definiteness is clear. For scalar multiplication, let $\vec{x} = x_1, ..., x_n$, and notice that

$$\|\alpha \cdot \vec{x}\|_2 = \sqrt{\sum_{i=1}^n (\alpha x_i)^2} = \sqrt{\alpha^2 \sum_{i=1}^n x_i^2} = |\alpha| \sqrt{\sum_{i=1}^n x_i^2} = |\alpha| \|\vec{x}\|_2.$$

Thus $\|\cdot\|_2$ is indeed a norm. We call $\|\cdot\|_2$ the **2-norm** or the **Euclidean** norm.

We observe that, in comparison with Example 9.1.3, we have that

$$d_2(\vec{x}, \vec{y}) = ||\vec{x} - \vec{y}||_2$$
.

Example 9.1.5 (1-norm)

Let $X = \mathbb{R}^n$, and $\vec{x} = (x_1, ..., x_n)$. Define

$$\|\vec{x}\|_1 := \sum_{i=1}^n |x_i|.$$

Clearly so, $\|\cdot\|_1$ is a norm:

- (positive definiteness) This is true by the absolute value function, i.e. every $|x_i| \ge 0$, and so the sum over these x_i 's is also nonnegative, and $\sum_{i=1}^n |x_i| = 0 \iff \forall i \in \{1,...,n\} \ x_i = 0 \iff \vec{x} = 0$.
- (scalar multiplication) This uses a similar argument as in the previous example.
- (triangle inequality) This is true by, again, the triangle inequality on absolute values.

We call $\|\cdot\|_1$ the **1-norm**.

Thus, we can define

$$d_1(\vec{x}, \vec{y}) = \|\vec{x} - \vec{y}\|_1$$

and it can easily be verified that d_1 is indeed a metric.

Example 9.1.6

Let $X = \mathbb{R}^n$ and $\vec{x} = (x_1, ..., x_n)$. Define

$$\|\vec{x}\|_{\infty} = \max\{|x_i|\}$$

Again, it is easy to see that $\|\cdot\|_{\infty}$ is a norm;

- (positive definiteness) : $\forall i \in \{1,...,n\} \ |x_i| \ge 0 \implies \max\{|x_i| \ge 0|$ and $\max\{|x_i|\}=0 \iff x_i=0 \iff \vec{x}=0.$
- (scalar multiplication) Notice that

$$\|\alpha \cdot \vec{x}\|_{\infty} = \max\{|\alpha x_i|\} = |\alpha| \max\{|x_i|\} = |\alpha| \|\vec{x}\|_{\infty}.$$

• (triangle inequality) This is once again true by the triangle inequality on the absolute value function, i.e.

$$\forall i \in \{1, ..., n\} \ |x_i + y_i| \le |x_i| + |y_i|$$

$$\max\{|x_i + y_i|\} \le \max\{|x_i|\} + \max\{|y_i|\}.$$

We can then define

$$d_{\infty}(\vec{x}, \vec{y}) = \max\{|x_i - y_i|\},\,$$

which we can easily verify that it is indeed a metric¹.

¹ Symmetry holds by the property of the absolute value function.

HERE's an interesting notion: let

$$S_i = \{ \vec{x} \in \mathbb{R}^2 \mid ||\vec{x}||_i = 1 \}, \quad i = 1, 2, \infty$$

Notice that we would then have the following graph: In fact, it is true

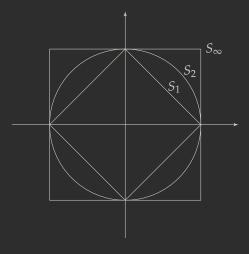


Figure 9.2: Unit ball depending on $\|\vec{x}\|_i$

that if we let $i \in \mathbb{N} \setminus \{0\}$, as suggested by Figure 9.2, we would see that the "diamond" would grow into a "circle" as in S_2 , and as $i \ge 3$, the unit ball will expand and approach the "square", which is S_{∞} .

Another observation that we can make is if we can show that a set

is open for a "smaller" S_i , then the same set is open for any S_j for j > i.

If we apply these norms into metrics, we have

$$d_{\infty} \leq d_2 \leq d_1$$

where we say that d_{∞} is the least sensitive, and d_1 being the most sensitive².

Example 9.1.7

For $1 , define on <math>\mathbb{R}^n$

$$\|\vec{x}\|_p = \left(\sum_{i=1}^n |x_i|^p\right)^{\frac{1}{p}}$$

Continuing with the same idea as in previous examples, we can let

$$d_p(\vec{x}, \vec{y}) = \left(\sum_{i=1}^n |x_i - y_i|^p\right)^{\frac{1}{p}}$$

In the next lecture, we will go into proving that this is indeed a norm, and so we can define a metric using this norm.

Note that if we allow for 0 < i < 1, then we would have a graph that looks like the following, which is a convex graph, i.e. we cannot create well-defined norms.

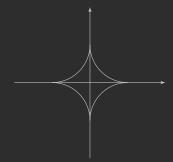


Figure 9.3: $\|\cdot\|_p$ for 0

 2 For sufficently close points, we see that d_∞ would reflect the least change, while we can see change in d_1 for every two points that we take.

10 Lecture 10 Sep 28th

10.1 Introduction to Metric Spaces (Continued)

Definition 33 ($\|\cdot\|_p$ -norm)

Given $\vec{x} = (x_1, ..., x_n) \in \mathbb{R}^n$, we define, for $1 , the <math>\|\cdot\|_p$ -norm to be

$$\left\|\vec{x}\right\|_{p} = \left(\sum_{i=1}^{n} |x_{i}|^{p}\right)^{\frac{1}{p}}$$

We asked the question: why is $\|\cdot\|_p$ a norm?

♣ Lemma 27

If
$$1 ,$$

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

and if α , beta > 0, then

$$\alpha \cdot \beta \le \frac{\alpha^p}{p} + \frac{\beta^q}{q}.$$

Proof

Motivated by Figure 10.1, using notions from calculus, we have from calculus,

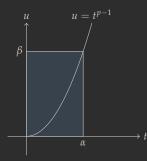


Figure 10.1: Motivation for Lemma 27.

$$\alpha\beta \le \int_0^\alpha t^{p-1} dt + \int_0^\beta u^{q-1} du$$
$$= \frac{t^p}{p} \Big|_0^\alpha + \frac{u^q}{q} \Big|_0^\beta$$
$$= \frac{\alpha^p}{p} + \frac{\beta^q}{q},$$

where we note that

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

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

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

$$1 = (p - 1)(q - 1)$$

Theorem 28 (Hölder's Inequality)

For $1 , let <math>\frac{1}{p} + \frac{1}{q} = 1$ ¹. Let

$$\vec{x} = (x_1, ..., x_n), \quad \vec{y} = (y_1, ..., y_n).$$

Then

$$\sum_{i=1}^{n} |x_i y_i| \le \left(\sum_{i=1}^{n} |x_i|^p \right)^{\frac{1}{p}} \cdot \left(\sum_{i=1}^{n} |y_i|^q \right)^{\frac{1}{q}}.$$

66 Note

Note that p = 2 *is just the Cauchy-Schwarz Inequality:*

$$\sum_{i=1}^{n} |x_i y_i| \le \left(\sum_{i=1}^{n} |x_i|^2\right)^{\frac{1}{2}} \cdot \left(\sum_{i=1}^{n} |y_i|^2\right)^{\frac{1}{2}} \Longrightarrow \left(\sum_{i=1}^{n} |x_i y_i|\right)^2 \le \left(\sum_{i=1}^{n} |x_i|^2\right) \cdot \left(\sum_{i=1}^{n} |y_i|^2\right)$$

¹ We also call q the conjugate of p.

Proof

Since if either \vec{x} or \vec{y} is zero, then we have that the inequality is trivially true, we can suppose that $\vec{x} \neq 0 \neq \vec{y}$. Now, note that for $\alpha, \beta \neq 0$, we have that²

$$\sum_{i=1}^{n} |x_i y_i| \le \left(\sum_{i=1}^{n} |x_i|^p\right)^{\frac{1}{p}} \cdot \left(\sum_{i=1}^{n} |y_i|^q\right)^{\frac{1}{q}}$$

$$\updownarrow$$

$$\sum_{i=1}^{n} |\alpha x_i \cdot \beta y_i| \le \left(\sum_{i=1}^{n} |\alpha x_i|^p\right)^{\frac{1}{p}} \cdot \left(\sum_{i=1}^{n} |\beta y_i|^q\right)^{\frac{1}{q}}.$$

So we can assume that

$$\left(\sum_{i=1}^{n} |x_i|^p\right)^{\frac{1}{p}} = 1 = \left(\sum_{i=1}^{n} |y_i|^q\right)^{\frac{1}{q}},\tag{10.1}$$

and if not, we can simply choose $\alpha, \beta \neq 0$ to scale these values to become one. By Lemma 27, we have

$$|x_iy_i| \leq \frac{|x_i|^p}{p} + \frac{|y_i|^q}{q}.$$

Hence

$$\sum_{i=1}^{n} |x_i y_i| \le \sum_{i=1}^{n} \frac{|x_i|^p}{p} + \sum_{i=1}^{n} \frac{|y_i|^q}{q} = \frac{1}{p} + \frac{1}{q} \quad \therefore Equation (10.1)$$

$$= 1 = \left(\sum_{i=1}^{n} |x_i|^p\right)^{\frac{1}{p}} \cdot \left(\sum_{i=1}^{n} |y_i|^q\right)^{\frac{1}{q}}$$

as required.

We are now ready to prove our long-awaited result.

Theorem 29 (Minkowski's Inequality)

Let
$$1 . If$$

$$\vec{x} = (x_1, ..., x_n), \quad \vec{y} = (y_1, ..., y_n)$$

² In the second inequality, notice that we can easily get back to the first equation by dividing both sides by $\alpha\beta$.

in \mathbb{R}^n , then

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

i.e.

$$\|\vec{x} + \vec{y}\|_{p} \le \|\vec{x}\|_{p} + \|\vec{y}\|_{p}.$$

Proof

Let

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

Once again, we may assume that $\vec{x} \neq 0 \neq \vec{y}$, as otherwise the inequality is true trivially so. Now, notice that

$$\sum_{i=1}^{n} |x_{i} + y_{i}|^{p} = \sum_{i=1}^{n} |x_{i} + y_{i}| |x_{i} + y_{i}|^{p-1}$$

$$\leq \sum_{i=1}^{n} |x_{i}| |x_{i} + y_{i}|^{p-1} + \sum_{i=1}^{n} |y_{i}| |x_{i} + y_{i}| \quad \text{triangle}$$

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

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

where the last step is by Hölder's Inequality. Note that $\frac{1}{p} + \frac{1}{q} = 1 \implies p = q(p-1)$. Thus

$$\sum_{i=1}^{n} |x_i + y_i|^p \le \left[\left(\sum_{i=1}^{n} |x_i|^p \right)^{\frac{1}{p}} + \left(\sum_{i=1}^{n} |y_i|^p \right)^{\frac{1}{p}} \right] \cdot \left(\sum_{i=1}^{n} |x_i + y_i|^p \right)^{\frac{1}{q}}$$

$$\implies \left(\sum_{i=1}^{n} |x_i + y_i|^p \right)^{1 - \frac{1}{q}} \le \left(\sum_{i=1}^{n} |x_i|^p \right)^{\frac{1}{p}} + \left(\sum_{i=1}^{n} |y_i|^p \right)^{\frac{1}{p}}$$

SS Note

With this we have that $\|\cdot\|_p$ satisfies the triangle inequaltiy condition, and so $\|\cdot\|_v$ is a norm on \mathbb{R}^n .

66 Note

Given $1 \le p \le q < \infty$, we have³

$$\|\cdot\|_{\infty} \le \|\cdot\|_{a} \le \|\cdot\|_{n} \le \|\cdot\|_{1}$$
.

³ For a visual representation of this result, see Figure 9.2.

Proof

It is quite clear that $\forall p \geq 1$,

$$\|\cdot\|_{\infty} = \max\{|\cdot|\} \le \left(\sum |\cdot|^{p}\right)^{\frac{1}{p}} = \|\cdot\|_{p}.$$

For $1 \le p \le q < \infty$, consider Holder's Inequality, where we have

$$\sum_{i=1}^{n} |a_i| |b_i| \le \left(\sum_{i=1}^{n} |a_i|^r\right)^{\frac{1}{r}} \cdot \left(\sum_{i=1}^{n} |b_i|^{\frac{r}{r-1}}\right)^{1-\frac{1}{r}}.$$

Let $|a_i| = |x_i|^p$, $|b_i| = 1$ and $r = \frac{q}{p} \ge 1$ 4. Then we have

⁴ Note that this is true by $p \le q$.

$$\sum_{i=1}^{n} |x_i|^p \le \left(\sum_{i=1}^{n} |x_i|^q\right)^{\frac{p}{q}} \cdot \left(\sum_{i=1}^{n} 1^{\frac{q}{q-p}}\right)^{1-\frac{p}{q}}$$
$$= n^{1-\frac{p}{q}} \cdot \left(\sum_{i=1}^{n} |x_i|^q\right)^{\frac{p}{q}}$$

Therefore, for $\vec{x} = (x_1..., x_n) \in \mathbb{R}$,

$$\begin{aligned} \|\vec{x}\|_{p} &= \left(\sum_{i=1}^{n} |x_{i}|^{p}\right)^{\frac{1}{p}} \leq \left(n^{1 - \frac{p}{q}} \cdot \left(\sum_{i=1}^{n} |x_{i}|^{q}\right)^{\frac{p}{q}}\right)^{\frac{1}{p}} \\ &= n^{\frac{1}{p} - \frac{1}{q}} \cdot \left(\sum_{i=1}^{n} |x_{i}|^{q}\right)^{\frac{1}{q}} = n^{\frac{1}{p} - \frac{1}{q}} \cdot \|\vec{x}\|_{q}. \end{aligned}$$

Thus, we have

$$\left\|\cdot\right\|_q \leq \left\|\cdot\right\|_p.$$

The chain of inequality follows.

1. Let
$$l_1 = \left\{ \left\{ x_i \right\} \mid \sum_{i=1}^{\infty} |x_i| < \infty \right\}$$
. Define

$$\|\{x_i\}\|_1 = \sum_{i=1}^{\infty} |x_i|$$

Let $\{x_i\}, \{y_i\} \in l_1$. Observe that $\forall n \in \mathbb{N}$, we have

$$\sum_{i=1}^{n} |x_i + y_i| \le \sum_{i=1}^{n} |x_i| + \sum_{i=1}^{n} |y_i| \le ||\{x_i\}||_1 + ||\{y_i\}||_1.$$

Then by the Monotone Convergence Theorem, we have that

$$\sum_{i=1}^{\infty} |x_i + y_i| \le \|\{x_i\}\|_1 + \|\{y_i\}\|_1.$$

Thus $\{x_i + y_i\} \in l_1$ and

$$\|\{x_i+y_i\}\|_1 \le \|\{x_i\}\|_1 + \|\{y_i\}\|_1$$
.

Let $\{x_n\} \in l_1$ and $\alpha \in \mathbb{R}$. Then

$$\sum_{i=1}^{\infty} |\alpha x_i| = |\alpha| \sum_{i=1}^{\infty} |x_i|.$$

Therefore $\{\alpha x_n\} \in l_1$ and $\|\{\alpha x_n\}\|_1 = |\alpha| \|\{x_n\}\|_1$.

Thus, we have that l_1 is a vector space, and $(l_1, \|\cdot\|_1)$ is a normed linear space.

2. Let $l_{\infty}(\mathbb{N}) = l_{\infty} = \{\{x_i\} \mid \{x_i\} \text{ is bounded }\}$. Define

$$\|\{x_i\}\|_{\infty} = \sup\{|x_1| \mid i \in \mathbb{N}\}.$$

Observe that $\forall \{x_i\}, \{y_i\} \in l_{\infty}$, then $\forall i \in \mathbb{N}$, we have

$$|x_i + y_i| \le |x_i| + |y_i| \le ||\{x_i\}||_{\infty} + ||\{y_i\}||_{\infty}.$$

So $\{x_i + y_i\} \in l_{\infty}$, and

$$\|\{x_i+y_i\}\|_{\infty} \leq \|\{x_i\}\|_{\infty} + \|\{y_i\}\|_{\infty}.$$

Consequently so, $\{\alpha x_i\} \in l_{\infty}$ and

$$\|\{\alpha x_i\}\|_{\infty} = |\alpha| \|\{x_i\}\|_{\infty}.$$

QUESTION: What about $l_p(\mathbb{R})$?

11 Lecture 11 Oct 01st

11.1 *Introduction to Metric Spaces (Continued 2)*

We wondered about $l_p(\mathbb{R})$ in the last lecture but let us consider a case that is even more general.

QUESTION: Can we define $l_p(\Gamma)$ for any set Γ ?

Example 11.1.1

Let
$$l_{\infty}(\Gamma) = \{ f : \Gamma \to \mathbb{R} \mid f(\Gamma) \text{ is bounded } \}$$
. If $f \in l_{\infty}(\Gamma)$, define

$$||f||_{\infty} = \sup\{|f(x)| \mid x \in \Gamma\}.$$

Notice that for f, $g \in l_{\infty}(\Gamma)$, and $\alpha \in \mathbb{R}$, then we have, by the Triangle Inequality,

$$||f + g||_{\infty} = \sup\{|(f + g)(x)| \mid x \in \Gamma\}$$

$$= \sup\{|f(x) + g(x)| \mid x \in \Gamma\}$$

$$\leq \sup\{|f(x)| \mid x \in \Gamma\} + \sup\{|g(x)| \mid x \in \Gamma\}$$

$$= ||f||_{\infty} ||g||_{\infty}.$$

So $f + g \in l_{\infty}(\Gamma)$, and

$$||f + g||_{\infty} \le ||f||_{\infty} + ||f||_{\infty}.$$

Also, we have

$$\begin{aligned} \|\alpha f\|_{\infty} &= \sup\{|(\alpha f)(x)| \mid x \in \Gamma\} \\ &= \sup\{|\alpha| \mid f(x)| \mid x \in \Gamma\} \\ &= |\alpha| \sup\{|f(x)| \mid x \in \Gamma\} \\ &= |\alpha| \mid \|f\|_{\infty}. \end{aligned}$$

So $\alpha f \in l_{\infty}(\Gamma)$, and $\|\alpha f\|_{\infty} = |\alpha| \|f\|$.

Therefore, $(l_{\infty}(\Gamma), \|\cdot\|_{\infty})$ is a normed linear space.

Example 11.1.2

Let $l_1(\Gamma) = \{f : \Gamma \to \mathbb{R} \mid P(f)\}$, where P(f) is the statement

$$||f||_1 = \sup \left\{ \sum_{i=1}^n |f(x_i)| \mid x_1, ..., x_n \in \Gamma, n \in \mathbb{N} \setminus \{0\} \right\} < \infty.$$

It is clear that $l_1(\Gamma) \subseteq l_{\infty}(\Gamma)$, where $l_{\infty}(\Gamma)$ is from Example 11.1.1. Consequently, $(l_1(\Gamma), \|\cdot\|_1)$ is a normed linear space.

We can extend the same idea onto l_p spaces.

Example 11.1.3

Let $X = C[a,b] = \{f : [a,b] \to \mathbb{R} \mid f \text{ is continuous on } [a,b]\}$, and define¹

$$||f||_{\infty} = \sup\{|f(x)| \mid x \in [a, b]\}$$

= \text{max}\{|f(x)| \crit x \in [a, b]\}

By (regular) Triangle Inequality, for any $f,g \in C[a,b]$, we have

$$||f + g||_{\infty} = \max\{|f(x) + g(x)| \mid x \in [a, b]\}$$

$$\leq \max\{|f(x)| \mid x \in [a, b]\} + \max\{|g(x)| \mid x \in [a, b]\}$$

$$= ||f||_{\infty} + ||g||_{\infty},$$

and, for $\alpha \in \mathbb{R}$,

$$\|\alpha f\|_{\infty} = \max\{|\alpha f(x)| \mid x \in [a, b]\}$$

= $|\alpha| \max\{|f(x)| \mid x \in [a, b]\}$
= $|\alpha| \|f\|_{\infty}$.

Thus $\|\cdot\|_{\infty}$ is a norm on C[a,b], and $(C[a,b],\|\cdot\|_{\infty})$ is a normed linear space^{2,3}.

Also, observe that

$$C[a,b] \subset l_{\infty}([a,b]).$$

Example 11.1.4

Let X = C[a, b] 4 have the same definition as the previous example,

Require clarification Notice that $\forall f \in l_1(\Gamma)$, for each $n \in \mathbb{N}$,

$$A_n = \{x \in \Gamma \mid |f(x)| \ge \frac{1}{n}\}$$
 is finite

So

$$A_0 = \bigcup_{n=1}^{\infty} A_n$$
 is countable.

and

$$A_0 = \{ x \in \Gamma \mid |f(x)| \neq \emptyset \}$$

¹ Note in this case sup is also max, since we are on a closed interval.

⁴ Some authors also write this as L'[a, b].

² This space is complete.

³ This space is important for us for the purpose of this course.

but this time define

$$||f||_1 = \int_a^b |f(t)| dt.$$

By linearity of integration, both the triangle equality and scalar multiplication hold, and so $(C[a,b], \|\cdot\|_1)$ is a normed linear space⁵.

⁵ Compared to the last example, this is not a complete space.

Example 11.1.5

Let X = C[a, b], and 1 . Define

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

Again, by linearity of integration, scalar multiplication holds. However, it is not as easy to show for the triangle inequality; we are now asking the same question as we did before for l_p , which we solved using Hölder's Inequality and Minkowski's Inequality. But now, instead of summations, we have integrations.

Theorem 30 (Hölder's Inequality v2)

Let $1 and <math>\frac{1}{n} + \frac{1}{a} = 1$. For each $f, g \in C[a, b]$, we have

$$\int_a^b |f(t)g(t)| dt \le \left(\int_a^b |f(t)|^p dt\right)^{\frac{1}{p}} \left(\int_a^b |g(t)|^q dt\right)^{\frac{1}{q}}.$$

Proof

If either f(x) = 0 or g(x) = 0 for all $x \in [a, b]$, then the inequality holds trivially so. Thus, we may assume that $\forall x \in [a, b], f(x) \neq$ $0 \neq g(x)$. By the linearity of integration, we can apply the same reasoning as we did in **P** Theorem 28, and assume that

$$\int_{a}^{b} |f(t)|^{p} dt = 1 = \int_{a}^{b} |g(t)|^{q} dt$$

By Lemma 27, we have

$$|f(t)g(t)| \le \frac{|f(t)|^p}{p} + \frac{|g(t)|^q}{q}.$$

Thus

$$\int_{a}^{b} |f(t)g(t)| dt \le \int_{a}^{b} \left(\frac{|f(t)|^{p}}{p} + \frac{|g(t)|^{q}}{q} \right) dt$$

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

$$= \left(\int_{a}^{b} |f(t)|^{p} dt \right)^{\frac{1}{p}} \left(\int_{a}^{b} |g(t)|^{q} dt \right)^{\frac{1}{q}}$$

as required.

Theorem 31 (Minkowski's Inequality v2)

Let $1 . If <math>f, g \in C[a, b]$, then

$$\left(\int_{a}^{b} |(f+g)(t)|^{p} dt\right)^{\frac{1}{p}} \leq \left(|f(t)|^{p} dt\right)^{\frac{1}{p}} \cdot \left(\int_{a}^{b} |g(t)|^{p} dt\right)^{\frac{1}{p}}.$$

Proof

The proof is similar to the one we had in \blacksquare Theorem 29; if $\forall x \in [a,b]$, either f(x)=0 or g(x)=0, then the inequality holds trivially so. Thus we may assyme that $\forall x \in [a,b]$, $f(x) \neq 0 \neq g(x)$. Now, notice that by (regular) Triangle Inequality and, later on,

Theorem 30,

$$\int_{a}^{b} |(f+g)(t)|^{p} dt
= \int_{a}^{b} |(f+g)(t)| |(f+g)(t)|^{p-1} dt
\leq \int_{a}^{b} |f(t)| |(f+g)(t)|^{p-1} dt + \int_{a}^{b} |g(t)| |(f+g)(t)| dt
\leq \left(\int_{a}^{b} |f(t)|^{p} dt\right)^{\frac{1}{p}} \left(\int_{a}^{b} |(f+g)(t)|^{q(p-1)} dt\right)^{\frac{1}{q}}
+ \left(\int_{a}^{b} |g(t)|^{p} dt\right)^{\frac{1}{p}} \left(\int_{a}^{b} |(f+g)(t)|^{q(p-1)} dt\right)^{\frac{1}{q}}
= \left[\left(\int_{a}^{b} |f(t)|^{p} dt\right)^{\frac{1}{p}} + \left(\int_{a}^{b} |g(t)|^{p} dt\right)^{\frac{1}{p}}\right]
\cdot \left(\int_{a}^{b} |(f+g)(t)|^{p}\right)^{\frac{1}{q}}$$

where we note that $\frac{1}{p} + \frac{1}{q} = 1 \implies p = q(p-1)$. Consequently,

$$\left(\int_{a}^{b} |(f+g)(t)|^{p} dt\right)^{\frac{1}{p}} \leq \left(|f(t)|^{p} dt\right)^{\frac{1}{p}} \cdot \left(\int_{a}^{b} |g(t)|^{p} dt\right)^{\frac{1}{p}},$$

as required.

This shows that our definition of $\|\cdot\|_p$ on C[a,b] is indeed a norm, and so $(C[a,b], \|\cdot\|_p)$ is a normed linear space.

Example 11.1.6 (Bounded Operator)

Let $(X, \|\cdot\|_X)$ and $(Y, \|\cdot\|_Y)$ be normed linear spaces. Let $T: X \to Y$ be linear. Define

$$||T|| = \sup\{||T_X||_Y \mid ||x||_X < 1\}.$$

We say that *T* is bounded if $||T|| < \infty$. Let

$$B(X,Y) = \{T : X \rightarrow Y \mid T \text{ is bounded } \}.$$

In the next lecture, we shall show that $(B(X,Y), \|\cdot\|)$ is a normed linear space.

QUESTION: Consider the transformation $\begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix} : \mathbb{R}^2 \to \mathbb{R}^2$. What is a norm $\begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix}$ that works?

Exercise 11.1.1

Show that there exists an injection from $(C[a,b], \|\cdot\|_2)$ to $l_2(\mathbb{N})$. Note that this does not work for $p \geq 3$.

A Useful Theorems from Earlier Calculus

■ Theorem 32 (Monotone Convergence Theorem)

Let $\{x_k\}$ be a sequence in \mathbb{R} .

- 1. Suppose $\{x_k\}$ is increasing.
 - If $\{x_k\}$ is bounded above, then $x_k \to \sup\{x_k\}$ as $k \to \infty$.
 - If $\{x_k\}$ is not bounded above, then $x_k \to \infty$ as $k \to \infty$.
- 2. Suppose $\{x_k\}$ is decreasing.
 - If $\{x_k\}$ is bounded below, then $x_k \to \inf\{x_k\}$ as $k \to \infty$.
 - If $\{x_k\}$ is not bounded below, then $x_k \to -\infty$.

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