

Lecture 5:

Zorn's Lemma

Let S be a partially ordered set. If every chain of S has an upper bound in S , then S contains a maximal elements.

Definition 1: (Partially ordered) A partially ordered on a (non-empty) set S is a binary relation on S , denoted \leq , which satisfies:

- for $\forall s \in S$, $s \leq s$
 - if $s \leq s'$ and $s' \leq s$, then $s = s'$
 - if $s \leq s'$ and $s' \leq s''$, then $s \leq s''$.
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Definition 2: If every elements in a partially ordered set S is comparable under \leq , then S is called a totally ordered set.

Definition 3: A chain is a collection of elements in S satisfying:
if $s_1 \in S$ and $s_2 \in S$, then either $s_1 \leq s_2$ or $s_2 \leq s_1$.

For simplicity, we may consider countable chain $s_1 \leq s_2 \leq \dots$ in our discussion
(not necessarily countable) for the ease of explanation

Definition 4: A maximal element m of a partially ordered set S is defined
as follows: for $\forall s \in S$ to which m is comparable, $s \leq m$.

Definition 5 An upper bound \tilde{C} of a chain $\{C_i\}_{i \in I}$ is that for all $i \in I$,
 $C_i \leq \tilde{C}$.

Zorn's lemma (adapted to 2048)

Let V be a vector space. Let S be the collection of linearly independent subsets of V . Then S is partially ordered under \subseteq .

Assume every chain $\{L_\alpha\}_{\alpha \in I}$ of S has an upper bound.

(Obviously, $\bigcup_{\alpha} L_{\alpha}$ is an upper bound of the chain. We need to show that

$\bigcup L_{\alpha}$ is linearly independent)

Then: S has a maximal element.

Theorem: Every vector space has a basis.

Proof: Let \mathcal{L} be the collection of all linearly independent subsets of V .

For any chain $\{S_i\}_{i \in I}$ (we may consider a countable chain $S_1 \subset S_2 \subset \dots$ for easier interpretation.)

$\bigcup_{i \in I} S_i$ is also linearly independent. $\therefore \bigcup_i S_i \in \mathcal{L}$.

By Zorn's lemma, \exists maximal linearly independent set M .

We claim that $\text{span}(M) = V$.

If not, $\exists \vec{v} \in V \ni \vec{v} \notin \text{span}(M)$.

Then: $M \cup \{\vec{v}\}$ is linearly independent.

But $M \subset M \cup \{\vec{v}\}$. Contradiction to Zorn's lemma.

\therefore ① $\text{span}(M) = V$

② M is L.I.

$\Rightarrow M$ is a basis.

Theorem: Every spanning set of a non-zero vector space V contains a basis of V .

Proof: Let S be a spanning set of V .

Let \mathcal{E} be the collection of linearly independent subsets of S .

Then $\mathcal{E} \neq \emptyset$ (as $\{\vec{v}\} \in \mathcal{E}$ for any $\vec{v} \neq \vec{0} \in S$)

Then, \mathcal{E} under \subseteq is partially ordered.

Let $\{L_i\}_{i \in I}$ be a chain in \mathcal{E} .

Then $\bigcup_i L_i \in \mathcal{E}$ and $\bigcup_i L_i$ is an upper bound.

By Zorn's lemma, there is a maximal element \mathcal{B} in \mathcal{E} .

(i.e. a linearly independent subset of S which is maximal)

We'll show that $\text{Span}(\mathcal{B}) = \text{Span}(\mathcal{S}) = V$

It suffices to show that for any $\vec{v} \in \mathcal{S}$, $\vec{v} \in \text{Span}(\mathcal{B})$.

If not, suppose $\vec{v} \notin \text{Span}(\mathcal{B})$.

Then: $\mathcal{B} \cup \{\vec{v}\}$ is linearly independent subset of \mathcal{S} .

Hence, $\mathcal{B} \cup \{\vec{v}\} \in \mathcal{L}$. But $\mathcal{B} \cup \{\vec{v}\} \supsetneq \mathcal{B}$.

Contradicting to the fact that \mathcal{B} is maximal.

$\therefore \vec{v} \in \text{Span}(\mathcal{B})$.

$\therefore \text{Span}(\mathcal{B}) = \text{Span}(\mathcal{S}) = V$.

$\therefore \mathcal{B}$ is a basis.

Remark: To find a basis inside a spanning set \mathcal{S} , it's natural to find a minimal spanning set of V inside \mathcal{S} .

If $M \subset \mathcal{S}$ is minimal, then M is linearly independent.

If not, we can find $v \in M \ni \text{Span}(M \setminus \{v\}) = \text{Span}(M)$, contradicting the minimality of M .

One might consider Zorn's lemma as follows:

Let \mathcal{C} be the set of all spanning subsets of \mathcal{S} , partially order \mathcal{C} by reverse inclusion. That is: $S_1 \in \mathcal{C}$ and $S_2 \in \mathcal{C}$, $S_1 \leq S_2$ iff $S_1 \supseteq S_2$.

For any chain $\{S_i\}_{i \in I}$ in \mathcal{C} , $\bigcap_{i \in I} S_i$ is the upper bound.

If $\bigcap_{i \in I} S_i \in \mathcal{C}$, then Zorn's lemma tells us \mathcal{C} has a maximal element (i.e. minimal spanning set)

BUT: $\bigcap_{i \in I} S_i$ MAY NOT always be in \mathcal{C} !!

Linear Transformation

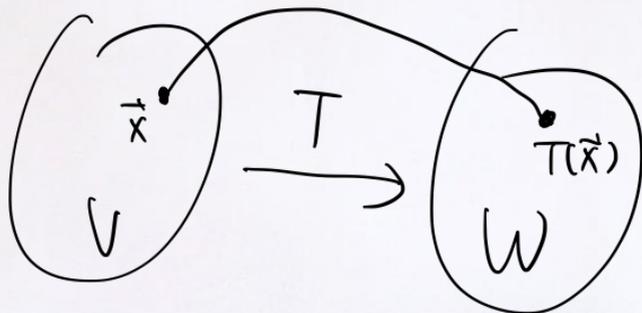
Definition: Let V and W be vector spaces over F .

A linear transformation from V to W is a map $T: V \rightarrow W$

such that: (a) $T(\vec{x} + \vec{y}) = T(\vec{x}) + T(\vec{y})$

(b) $T(a\vec{x}) = aT(\vec{x})$

for all $\vec{x}, \vec{y} \in V$, $a \in F$.



Proposition: Let $T: V \rightarrow W$ be a linear transformation. Then:

$$(i) \quad T(\vec{0}_V) = \vec{0}_W$$

$$(ii) \quad T\left(\sum_{i=1}^n a_i \vec{x}_i\right) = \sum_{i=1}^n a_i T(\vec{x}_i) \quad \forall \vec{x}_1, \vec{x}_2, \dots, \vec{x}_n \in V$$

(T preserves linear combination) $a_1, a_2, \dots, a_n \in F.$

$$(i) \quad T(\vec{0}_V) = T(\vec{0}_V + \vec{0}_V) = T(\vec{0}_V) + T(\vec{0}_V)$$

$$\Rightarrow T(\vec{0}_V) = \vec{0}_W. \quad (\text{Cancellation law})$$

(ii) Use math. induction (exercise)

Examples: • For any vector spaces V and W , we have:

(a) The **zero transformation** $T_0: V \rightarrow W$ defined by $T_0(\vec{x}) := \vec{0}_W$
for $\forall \vec{x} \in V$

(b) The **identity transformation** $I_V: V \rightarrow V$ defined by $I_V(\vec{x}) = \vec{x}$
for $\forall \vec{x} \in V$.

• Let $A \in M_{m \times n}(F)$ be a $m \times n$ matrix. F .

Define: $L_A: F^n \rightarrow F^m$ as: ($F^n =$ space of col vectors of size n)

$$L_A(\vec{x}) \stackrel{\text{def}}{=} A\vec{x}$$

L_A is called the left multiplication by A .

• $T: M_{m \times n}(F) \rightarrow M_{n \times m}(F)$ defined by $T(A) \stackrel{\text{def}}{=} A^t$ (transpose of A)

- $T: P_n(\mathbb{R}) \rightarrow P_{n-1}(\mathbb{R})$ defined by $T(f(x)) = f'(x)$
is a lin. transf. (derivative of f)

- Let a and $b \in \mathbb{R}$, $a < b$. Then,

$T: C(\mathbb{R}) \rightarrow \mathbb{R}$ defined by =
(Space of continuous functions)

$$T(f) \stackrel{\text{def}}{=} \int_a^b f(t) dt$$

Null space or Range

Definition: Let V and W be vector spaces and $T: V \rightarrow W$ be a linear transformation

Then, the null space (or kernel) of T is defined as:

$$N(T) := \text{def } \{ \vec{x} \in V : T(\vec{x}) = \vec{0}_W \} \subset V$$

the range (or image) of T is defined as:

$$R(T) := \{ T(\vec{x}) : \vec{x} \in V \} \subset W$$

e.g. For $I_V: V \rightarrow V$, $N(I_V) = \{ \vec{0}_V \}$, $R(I_V) = V$
(identity)

For $T_0: V \rightarrow W$, $N(T_0) = V$, $R(T_0) = \{ \vec{0}_W \}$
(zero transf)

• $L_A: F^n \rightarrow F^m$ ($A \in M_{m \times n}(F)$)

$N(L_A) = N(A) =$ null space of A

$R(L_A) = \mathcal{C}(A) =$ col space of A

(space of linear combination of col vectors of A)

• For $T: P_n(\mathbb{R}) \rightarrow P_{n-1}(\mathbb{R})$ defined by

$T(f(x)) = f'(x)$, then:

$N(T) = \{ a_0 \in P_n(\mathbb{R}) : a_0 \in \mathbb{R} \}$

$R(T) = P_{n-1}(\mathbb{R})$

$A = \left(\begin{array}{c|c|c|c} | & | & \dots & | \end{array} \right)$