

Math 146 Notes

velo.x

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Section: 001

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1 Vector Space

1.1 Vector Space Jan 6

Definition 1.1.1 (Pseudo-Field). A field is an algebraic system \mathbb{F} having:

- two elements 0 and 1
- operations $+$, \times , $-$, and $()^{-1}$ (defined on nonzero elements)

satisfying "the obvious" properties.

See appendix of the textbook.

Examples: \mathbb{R} , \mathbb{C} , \mathbb{Q} , $\mathbb{Z}_{\text{prime}}$. $\mathbb{Q}(x) = \left\{ \frac{f(x)}{g(x)} : f, g \text{ polynomials}, g \neq 0 \right\}$

NonExamples: $\{0\}$, \mathbb{Z}_m (m not prime), Quaternions.

Definition 1.1.2 (Vector Space). A vector space over \mathbb{F} is a set V with two operations:

- Addition: $V \times V \rightarrow V$ $x + y$
- Scalar Multiplication: $\mathbb{F} \times V \rightarrow V$ ax

satisfying 8 properties: $\forall x, y, z \in V, \forall a, b \in \mathbb{F}$

- V1: $x + y = y + x$
- V2: $x + (y + z) = (x + y) + z$
- V3: \exists a "zero vector" $0 \in V$ s.t. $x + 0 = x$
- V4: $\forall x \in V, \exists u \in V$, s.t. $x + u = 0$
- V5: $1x = x$
- V6: $(ab)x = a(bx)$ *let \cdot denote scalar multiplication
- V7: $a(x + y) = ax + ay$
- V8: $(a + b)x = ax + bx$

Objective 1.1.1.

- Defining/Constructing
- Proving that a system is a vector space

Example 1: \mathbb{R} def: set of all n – tuples of real numbers

$$(x_1, \dots, x_n) + (y_1, \dots, y_n) =$$

$a(x_1, \dots, x_n)$ defined (ax_1, \dots, ax_n) Claim: \mathbb{R}^n is a vector space over \mathbb{R}

Proof. Check V1:

$$\begin{aligned}(x_1, \dots, x_n) + (y_1, \dots, y_n) &= (x_1 + y_1, \dots, x_n + y_n) \\ &= (y_1 + x_1, \dots, y_n + x_n) \\ &= (y_1, \dots, y_n) + (x_1, \dots, x_n)\end{aligned}$$

□

More generally, for any field \mathbb{F} , \mathbb{F}^n is a field over \mathbb{F} .

Example 2: $\mathbb{R}^{[0,1]} = \{all\ functions\ f : [0, 1] \rightarrow \mathbb{R}\}$

- $(f + h)(x) \stackrel{def}{=} f(x) + g(x)$
- $(af)(x) = af(x)$

Claim: $\mathbb{R}^{[0,1]}$ is a vector space $/\mathbb{R}$.

Proof. V3: Let $\bar{0}$ be the constant 0 function, i.e., $\bar{0}(x) = 0 \ \forall x \in [0, 1]$ $\bar{0} \in \mathbb{R}^{[0,1]}$

Check: $f + \bar{0} = f \ \forall f \in \mathbb{R}^{[0,1]}$

$$\begin{aligned}(f + \bar{0})(x) &= f(x) + \bar{0}(x) \\ &= f(x) + 0 = f(x)\end{aligned}$$

Since $x \in [0, 1]$ arbitrary, $f + \bar{0} = f$.

More generally, for any set D, and any field \mathbb{F} , \mathbb{F}^D is a vector space over \mathbb{F} .

□

Example 3: let $\mathbb{F} = \mathbb{Z}_2$.

Define $W = \{APPLE\}$,

- $APPLE + APPLE \stackrel{def}{=} APPLE$
- $0APPLE \stackrel{def}{=} APPLE$
- $1APPLE \stackrel{def}{=} APPLE$

Claim: W is a vector space over \mathbb{Z}_2 .

1.2 Vector Space and Introduction to Linear Combination Jan 8

Examples: 1. $\mathbb{R}^n : \mathbb{F}^n$, 2. $\mathbb{R}^{[0,1]} : \mathbb{F}^D$, 3. $\{APPLE\}$.

4. Fix a field \mathbb{F} , for $n \geq 0$, $P_n(\mathbb{F})$ is the set of all polynomials, of degree $\leq n$, in variable x , with coefficients from \mathbb{F} ,

$$= \{a_0 + a_1x + a_2x^2 + \cdots + a_nx^n : a_i \in \mathbb{F}\}$$

Addition, scalar mult are "obvious", using op's of \mathbb{F} .

Claim: $P_n(\mathbb{F})$ is a vector space / \mathbb{F} .

5. $\mathbb{F}[x]$ = the set of all polynomials in x with coefficients from $\mathbb{F} = \cup_{n=0}^{\infty} P_n(\mathbb{F})$

Claim: with the "obvious" op's $\mathbb{F}[x]$ is a V.S. / \mathbb{F} .

Theorem 1.2.1 (Cancellation Law). *Let V be a V.S., / \mathbb{F} , if $x, y, z \in V$, and $x + z = y + z$, then $x = y$.*

Proof. Let $u \in V$ be such that $z + u = 0$ (from V4).

Then

$$\begin{aligned} x &= x + 0 && \text{(V3)} \\ x &= x + (z + u) && \text{(Choice of u)} \\ x &= (x + z) + u && \text{(hypothesis)} \\ x &= (y + z) + u && \text{(V2)} \\ x &= y + (z + u) && \text{(V2)} \\ x &= y + 0 && \text{(choice of u)} \\ x &= y \end{aligned}$$

□

Corollary 1.2.1. *Suppose V is a V.S., there is exactly one "zero vector". i.e. a vector satisfy V3. in V .*

Proof. Assume $0_1, 0_2 \in V$, both satisfying V3, i.e, $x + 0_1 = x$ and $x + 0_2 = x, \forall x \in V$.

$$\begin{aligned} 0_1 &= 0_1 + 0_1 \\ 0_1 &= 0_1 + 0_2 \\ 0_1 + 0_1 &= 0_1 + 0_2 \\ &= 0_2 + 0_1 && \text{(V1)} \\ 0_1 &= 0_2 && \text{(By Cancellation)} \end{aligned}$$

□

Corollary 1.2.2. *Suppose V is a V.S. and $x \in V$, then the vector u in V4 is unique.*

Proof. Assume $u_1, u_2 \in V$ both satisfy $x + u_1 = 0 = x + u_2$, then

$$u_1 + x = u_2 + x \quad (\text{V1})$$

$$u_1 = u_2 \quad (\text{By Cancellation})$$

□

Definition 1.2.1. Given a V.S. V and $x \in V$,

- the unique vector $u \in V$ s.t. $x + u = 0$ is denoted $-x$.
- $x - y$ denotes $x + (-y)$

Note: V2 justifies $a_1x_1 + a_2x_2 + a_3x_3 + \cdots + a_kx_k$ not worry about parentheses.

Definition 1.2.2 (Linear Combination). $a_1x_1 + a_2x_2 + a_3x_3 + \cdots + a_kx_k$ is called a linear combination of x_1, \cdots, x_k .

Basic Problem: Given a V.S. V/\mathbb{F} , and $u_1, u_2, \cdots, u_n \in V$ and $x \in V$ to decide whether x is a linear combination of u_1, \cdots, u_n .

Example: $V = \mathbb{Q}[x]$ over \mathbb{Q} . Let $p = 4x^4 + 7x^2 - 2x + 3$.

- $u_1 = x^4 - x^2 + 2x + 1$
- $u_2 = 2x^4 + 3x^2 + 2x$
- $u_3 = x^4 + 4x^2 + 1$
- $u_4 = 2x^3 + 3$
- $u_5 = x^4 + 1$

Is p a linear combination of u_1, \cdots, u_5 ? **Solution:** search for $a_1, \cdots, a_5 \in \mathbb{Q}$ s.t.

$$p = a_1u_1 + a_2u_2 + \cdots + a_5u_5$$

$$\begin{aligned} 4x^4 + 7x^2 - 2x + 3 &= a_1(x^4 - x^2 + 2x - 1) + a_2(2x^4 + 3x^2 + 2x) + a_3(x^4 + 4x^2 + 1) \\ &\quad + a_4(2x^3 + 3) + a_5(x^4 + 1) \\ &= (a_1 + 2a_2 + a_3 + a_5)x^4 + (2a_4)x^3 + (-a_1 + 3a_2 + 4a_3)x^2 \\ &\quad + (2a_1 + 2a_2)x + (-a_1 + a_3 + 3a_4 + a_5) \end{aligned}$$

$$\begin{cases} a_1 + 2a_2 + a_3 + a_5 = 4 \\ 2a_4 = 0 \\ -a_1 + 3a_2 + 4a_3 = 7 \\ 2a_1 + 2a_2 = -2 \\ -a_1 + a_3 + 3a_4 + a_5 = 3 \end{cases}$$

No solution.

1.3 Subspace Jan 10

Notation 1.3.1.

- 0 denote the unique vector in V
- x denote the unique $u \in V$ satisfying $V4$

Theorem 1.3.1. Suppose V is a VS/ \mathbb{F} , $X \in V$, $a \in \mathbb{F}$.

1. $0x=0$, the first 0 is scalar, the second 0 is a vector
2. $(-a)x=a(-x)=- (ax)$
3. $a0=0$

Definition 1.3.1. Suppose V is a V.S. over \mathbb{F} , $S \subseteq V$,

- **Closed under Addition:** if $x, y \in S$, $x + y \in S$.
- **Closed under Scalar Multiplication:** if $x \in S \Rightarrow ax \in S$, $\forall a \in \mathbb{F}$.

Definition 1.3.2 (Subspace). Let V be a VS/ \mathbb{F} , $S \subseteq V$, say S is a **Subspace** of V if

1. S is closed under addition and scalar multiplication
2. $S \neq \emptyset$

Theorem 1.3.2. Suppose V is a vector space / \mathbb{F} and S is a subspace of V , then S , together the operations of V restricted to S .

- $+_S : S \times S \rightarrow S$
- $\cdot_S : \mathbb{F} \times S \rightarrow S$

Proof. Given V, S , must prove: S with restricted operations of V , satisfying $V1$ to $V8$.

V1: must show: if $x, y \in S$, then $x + y = y + x$. Since $S \subseteq V$, hence $x, y \in S \Rightarrow x, y \in V$, and $V \models V1$.

Same proof works for $V2, 5, 6, 7, 8$.

For $V3$, know $S \neq \emptyset$, take any $x \in S$, consider $0x = 0 \in S$. (S is closed under scalar multiplication)

Hence there exists a zero vector in S .

For $V4$, fix $x \in S$, let $u = (-x)x \in S$, then $x + u = 1x + (-1)x = (1 + (-1))x = 0x = 0$. □

Note: in every \mathbb{F} , $\forall a \in \mathbb{F}$, $\exists c \in \mathbb{F} a + c = 0$, $c = -a$. Since $1 \in \mathbb{F}$, $-1 \in \mathbb{F}$.

Theorem 1.3.3. If V is a vector space over \mathbb{F} and $S \subseteq V$, and S with the operations of V , is itself a V.S. / \mathbb{F} , then S is a subspace of V .

1.4 Span Jan 13

Recall: If V is a V.S. / \mathbb{F} , and $u_1, \dots, u_n, x \in V$, then x is a linear combination (lin. combo.) of u_1, \dots, u_n if $\exists a_1, \dots, a_n$ such that $x = a_1u_1 + a_2u_2 + \dots + a_nu_n$.

Definition 1.4.1. Suppose V is a V.S. / \mathbb{F} , $x \in V$, and $\emptyset \neq S \subseteq V$.

1. Say x is a lin. combo. of S if \exists finitely many $u_1, \dots, u_n \in S$, s.t. x is a lin. combo. of u_1, \dots, u_n .
 $S = \{u_1, u_2, \dots, u_n\}$, $x = \sum_{n=0}^{\infty} a_n u_n$, converge.
2. The **Span** of S written $\text{span}(S)$, is the set of all linear combinations of S .
3. $\text{span}(\emptyset) \stackrel{\text{df}}{=} \{0\}$

Examples

- In \mathbb{R}^2 , $S = \{(1, 1)\}$, what is $\text{span}(S)$? the
- In \mathbb{R}^3 , $S = \{(1, 0, 0), (1, 1, 0)\} = \{a(1, 0, 0) + b(1, 1, 0) : a, b \in \mathbb{R}\} = \{(a + b, b, 0) : a, b \in \mathbb{R}\} = (s, t, 0) : s, t \in \mathbb{R}$ = the plane given by $z = 0$
- In $\mathbb{R}[x]$, let $S = \{x, x^2, x^3, \dots\}$, $\text{span}(S) = \{f \in \mathbb{R}[x] : f(0) = 0\}$.

Proposition 1.4.1. ($\emptyset \neq S \subseteq V$). Suppose $u_1, \dots, u_n \in S$, $x \in V$. Suppose x is a linear combination of u_1, \dots, u_n . If v_1, \dots, v_n are more vectors from S , then x is also a linear combination of $u_1, \dots, u_n, v_1, \dots, v_n$.

Proposition 1.4.2. If $S = \{u_1, \dots, u_n\}$, then $\text{span}(S) = \{a_1u_1, \dots, a_ku_k, a_1, \dots, a_k \in \mathbb{F}\}$.

Proposition 1.4.3. If $S \subseteq T \subseteq V$, then $\text{span}(S) \subseteq \text{span}(T)$.

Proposition 1.4.4. If S is infinite, if $x, y \in \text{span}(S)$, say x is a linear combo of $u_1, \dots, u_n \in S$, y is a linear combo of $v_1, \dots, v_n \in S$, then x, y are linear combos of $u_1, \dots, u_n, v_1, \dots, v_n$.

Generalization 1.4.1. If $x_1, \dots, x_k \in \text{span}(S)$, then $\exists u_1, \dots, u_n \in S$, s.t. each x_l is a linear combo of u_1, \dots, u_n .

Theorem 1.4.1. Suppose V is a V.S. / \mathbb{F} , $S \subseteq V$, then $\text{span}(S)$ is the (unique) smallest subspace of $V \supseteq S$. i.e.

1. $\text{span}(S)$ is a subspace of V .
2. $S \subseteq \text{span}(S)$
3. If W is any subspace of V containing S , then $\text{span}(S) \subseteq W$.

Proof. (2) Let $x \in S$, $x = 1x$, a linear combination of finitely many vectors in S .

(1) i) Closure under scalar multiplication: let $x \in \text{span}(S)$, $c \in \mathbb{F}$, $\Rightarrow \exists u_1, \dots, u_n \in S$, s.t. $x = a_1x_1 + \dots + a_nx_n$, so

$$cx = c(a_1u_1 + \dots + a_nu_n) = (ca_1)u_1 + \dots + (ca_n)u_n$$

II) Closure under vector addition: let $x, y \in \text{span}(S)$, want to prove that $x + y \in \text{span}(S)$.

By the technical remark, $\exists u_1, \dots, u_n \in S$ s.t. $x = a_1u_1 + \dots + a_nu_n$, $y = b_1u_1 + \dots + b_nu_n$, $a_i, b_i \in \mathbb{F}$,

Then, $x + y = (a_1u_1 + \dots + a_nu_n) + (b_1u_1 + \dots + b_nu_n) = (a_1 + b_1)u_1 + \dots + (a_n + b_n)u_n$.

So $x + y \in \text{span}(S)$.

Finally, if $S = \emptyset$, then $\text{span}(S) = \{0\}$, if $S \neq \emptyset$, then $S \subseteq \text{span}(S)$,

either case, $\text{span}(S) \neq \emptyset$, so $\text{span}(S)$ is a subspace of V .

3) Let W be a subspace

□

Intuition: Redundancies in span. Example: V / \mathbb{F} , suppose $S = \{u_1, \dots, u_5\} \subseteq V$.

Assume u_3 is a linear combination of u_2, u_4, u_5 .

$$u_3 = c_2u_2 + c_4u_4 + c_5u_5$$

Claim: $\text{span}(S) = \text{span}(S - \{u_3\})$.

Proof. RTP \subseteq and \supseteq .

$\text{span}(S)$ is

- a subspace of V
- which contains $S \setminus \{u_3\} = \{u_1, u_2, \dots, u_5\}$

By the theorem, the smallest subspace of V containing $S \setminus \{u_3\}$ is $\text{span}(S \setminus \{u_3\})$. hence $\text{span}(S) \supseteq \text{span}(S \setminus \{u_3\})$.

To prove that $\text{span}(S) \subseteq \text{span}(S \setminus \{u_3\})$,

let $x \in \text{span}(S)$, i.e.

$$\begin{aligned} x &= a_1u_1 + a_2u_2 + a_3u_3 + a_4u_4 + a_5u_5 \\ &= a_1u_1 + a_2u_2 + a_3(c_2u_2 + c_4u_4 + c_5u_5) + a_4u_4 + a_5u_5 \\ &= a_1u_1 + (a_2 + a_3c_2)u_2 + (a_4 + a_3c_4)u_4 + (a_5 + a_3c_5)u_5 \end{aligned}$$

$x \in \text{span}(\{u_1, u_2, u_4, u_5\})$

□

Also Observe:

$$0u_1 + c_2u_2 + (-1)u_3 + c_4u_4 + c_5u_5 = 0$$

A linear combination of u_1, \dots, u_5 equals the 0 vector with coefficients not all 0.

So we code redundancies formally with definition:

Definition 1.4.2. ($V\mathbb{F}, S \subseteq V$), S is linearly dependent if \exists distinct vectors $u_1, \dots, u_n \in S$, and $\exists a_1, \dots, a_n \in \mathbb{F}$, not all 0, such that

$$a_1 u_1 + a_2 u_2 + \dots + a_n u_n = 0 (\text{zero vector})$$

S is linearly independent if S is not linearly dependent.

S is linearly dependent $\iff (\exists \text{ distinct } u_1, \dots, u_n \in S)(\exists a_1, \dots, a_n \in \mathbb{F}, \not\equiv 0)(a_1 u_1 + \dots + a_n u_n) = 0$
 $\equiv (\forall \text{ distinct } u_1, \dots, u_n \in S)(\quad)$

Technical Remark: when $S = \{u_1, \dots, u_n\}$ without reports

- Can drop $(\forall \text{ distinct } u_1, \dots, u_n \in S)$ in choice of linear independence.

-Can drop $(\exists \text{ distinct } u_1, \dots, u_n \in S)$ in choice of linear dependence.

Example 2: Is $S = \{(0, 1, 1), (1, 0, 1), (1, 2, 3)\}$ linear dependent? (in \mathbb{R}^3)

Try to find: $a, b, c \in \mathbb{R}$ s.t.

$$a \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} + b \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} + c \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \quad \Rightarrow \quad \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} -2 \\ -1 \\ 1 \end{pmatrix}$$

Shows S is linearly dependent.

Question: If $S = \emptyset$, S is linearly dependent.

Question 2: If $S = \{0\}$, S linearly dependent. Can write $1 \cdot 0 = 0$.

More Generally, if $0 \in S \subseteq V$, then S is linearly dependent.

Theorem 1.4.2 (Linear Dependence). $V\mathbb{F}, S \subseteq V$, then S is linearly dependent, iff $S = \{0\}$ or $\exists x \in S$, s.t. x is a linear combination of some vectors in $S \setminus \{x\}$.

1.5 Basis Jan 17

Recall If V is a V.S. / \mathbb{F} , $S \subseteq B$.

1. $\text{span}(S)$ = set of all linear combinations of S
2. S is linearly dependent if $\exists u_1, u_2, \dots, u_n \in S$ (distinct), $\exists a_1, \dots, a_n \in \mathbb{F}$ not all 0, s.t, $a_1 u_1 + a_2 u_2 + \dots + a_n u_n = 0$.
 -else, S is linearly independent.

Definition 1.5.1. V is V.S. / \mathbb{F} ,

1. A set $S \subseteq V$ is a spanning set of $\text{Span}(S) = V$. Also say S spans V .
2. V is finitely spanned if V has a finite spanning set.
 V is countably spanned if V has a countable spanning set.

Examples: \mathbb{R}^3 is finitely spanned, e.g. by $\{e_1, e_2, e_3\}$.

so is \mathbb{R}^n e.g. by $\{e_1, e_2, \dots, e_n\}$, $e_i = (0, 0, \dots, 1, 0, \dots, 0)$ with 1 at i_{th} spot.

$\mathbb{R}[x]$ is countably spanned e.g. by $\{1, x, x^2, x^3, \dots\}$. not finitely spanned.

$\mathbb{R}[0, 1]$ not countably spanned.

Definition 1.5.2. V is a V.S. / \mathbb{F} .

A basis for V is any $S \subseteq V$, which

- spans V , and
- S is linearly independent

Examples: $\{e_1, \dots, e_n\} \subseteq \mathbb{F}^n$ is a basis for \mathbb{F}^n .

$\{1, x, x^2, x^3, \dots\} \subseteq \mathbb{R}[x]$ is a basis for $\mathbb{R}[x]$.

Theorem 1.5.1. Every countably spanned V.S. has a basis.

Proof. Suppose V.S. V is spanned by countable set S , so either $S = \{v_1, v_2, \dots, v_n\}$, or $S = \{v_1, v_2, \dots\}$, WLOG, we assume $0 \notin S$, define

$$T = \{v_j \in S, v_j \notin \text{span}(v_1, v_2, \dots, v_{j-1})\},$$

Claim that T is a basis for V .

Proof of Claim: 1st show T is linearly independent, by contradiction, assume T is linearly dependent.

Then, $\exists k$, and scalars a_1, a_2, \dots, a_n (not all 0), s.t,

$$a_1x_1 + a_2x_2 + \dots + a_nx_n = 0$$

Choose least k for which this is true.

Claim: $k \neq 1$, if $k = 1$, $a_1v_1 = 0 \Rightarrow v_1 = 0$, but $0 \notin T$, contradiction.

so $k > 1$, Assume $a_k = 0$, then

$$a_1v_1 + a_2v_2 + \dots + a_{k-1}v_{k-1} = 0$$

Not all of $a_1, a_2, \dots, a_{k-1} = 0$.

Next, show $\text{span}(S)$

□

Remark:

1. Every Vector Space has a basis. proof: some version of axiom of choice
2. bases is not unique, every V.S. except $\{0\}$, has multiple bases.
3. What is a basis for $V = \{0\}$? \emptyset

Theorem 1.5.2 (Axiom of Choice). Suppose A, B are sets, $f : A \rightarrow$.

1.6 Dimensioning a Vector Space - Jan 20

Given a vector space V , Is a basis unique?

No.

Relation between two basis of a vector space. (finitely spanned vector spaces)

Theorem 1.6.1. *Let V be a finitely spanned vector space over a field \mathbb{F} , let $\{v_1, \dots, v_m\}$ be a basis of V , let $\{w_1, \dots, w_n\} \subset V$ and $n > m$. Then $\{w_1, \dots, w_n\}$ is linearly dependent.*

Sketch. Idea: Replace successfully v_1, v_2, \dots, v_n by w_1, w_2, \dots, w_n so that

$$\text{span}(\{w_1, w_2, \dots, w_i, v_{i+1}, \dots, v_m\}) = \text{span}(\{v_1, v_2, \dots, v_i, v_{i+1}\})$$

.

$$1 \leq i \leq m-1.$$

□

Proof. Assume $\{w_1, \dots, w_n\}$ is linearly independent. Prove the statement by induction.

Base Case: (i=1), since $\{v_1, \dots, v_m\}$ is a basis for V and $w_1 \in V$, there exist $a_1, \dots, a_m \in \mathbb{F}$ s.t. $w_1 = a_1 v_1 + \dots + a_m v_m$.

By the assumption, $w_1 \neq 0$, hence one of the a'_k s is nonzero.

By renumbering v_1, \dots, v_m , WLOG, we can assume $a_1 \neq 0$. We can solve for v_1 .

$$\begin{aligned} a_1 v_1 &= w_1 - a_2 v_2 - \dots - a_m v_m \\ v_1 &= a_1^{-1} w_1 - a_1^{-1} a_2 v_2 - \dots - a_1^{-1} a_m v_m \end{aligned}$$

so, $\text{span}(\{v_1, v_2, \dots, v_m\}) \subset \text{span}(\{w_1, w_2, \dots, w_m\}) = V$.

Induction Assumption: Assume that the statement is true for r . It means after renumbering, v_1, v_2, \dots, v_m we have

$$\text{span}(\{w_1, w_2, \dots, w_i, v_{i+1}, \dots, v_m\}) = V.$$

*replace w_{i+1} .

Prove for $r+1$: Rewrite w_{i+1} as a linear combination of $\{w_1, \dots, w_r, v_{r+1}, \dots, v_m\}$.

$$w_{i+1} = c_1 w_1 + \dots + c_r w_r + d_{i+1} v_{i+1} + \dots + d_m v_m$$

Observation: One of the d_{r+1}, \dots, d_m must be nonzero. Because if $d_{i+1} = \dots = d_m = 0$, then

$$\begin{aligned} w_{r+1} &= c_1 w_1 + \dots + c_r w_r \\ 0 &= c_1 w_1 + \dots + c_r w_r - w_{r+1} \end{aligned}$$

Contradiction since $\{w_1, \dots, w_{r+1}\}$ is linearly independent.

WLOG, we can assume $d_{i+1} \neq 0$,

$$d_{r+1} v_{r+1} = w_{r+1} - c_1 w_1 - \dots - c_r w_r - d_{r+2} v_{r+2} - \dots - d_m v_m$$

Since $n > m$, $w_n = a_i w_i + \dots + a_m w_m$, so $\{w_1, \dots, w_n\}$ is linearly dependent.

It completes the proof.

□

Theorem 1.6.2. Let V be a finitely spanned vector space, having one basis of m elements having another basis of n elements.

Then $m = n$.

Proof. We could not have $m < n$, or $m > n$. If it happens, the other set must be linearly dependent. \square

Definition 1.6.1. Let V be a vector space having a basis consisting of n elements, we say n is the dimensioning of V .

$$\dim_{\mathbb{F}} V = n$$

$$\lim\{0 = 0\}$$

A vector space that has a basis consisting of n elements, zero elements, zero vector space, is called finite dimensional. Otherwise, V is called infinite dimensional([Hamel Basis](#))

Example:

- $\dim \mathbb{F}^n = n$

Since

$$\left\{ \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \dots, \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \right\}$$

is a basis for \mathbb{F}^n .

- $\dim P_n(\mathbb{F}) = n + 1$

Since $\{1, x, \dots, x^n\}$ is a basis for $P_n(\mathbb{F})$.

- $\dim \mathbb{F}[x] = \infty$

Corollary 1.6.1. Let V be an n -dimensional space, then

- If $\{v_1, \dots, v_n\} \subset V$ is linearly independent, then $\{v_1, \dots, v_n\}$ is a basis for V .
- If $\{v_1, \dots, v_n\} \subset V$, $k < n$ is linearly we can add v_{k+1}, \dots, v_n so that $\{v_1, \dots, v_n\}$ is a basis for V .
- If W is a subspace of V , then $\dim W \leq \dim V$, if furthermore, $\dim W = \dim V$. Then $W = V$.

1.7 Direct Sum:

Corollary 1.7.1. *If V is finitely spanned, and $\beta\{v_1, \dots, v_n\}$ is linearly independent, then β can be extended to a basis for V , i.e. $\exists w_1, \dots, w_r \in V$, s.t. $\{v_1, \dots, v_n, w_1, \dots, w_r\}$ is a basis for V*

Proof. Let $m = \dim V$. So $n \leq m$ by theorem.

Case 1: β is already a basis. ($n=m$)

Case 2: β is not a basis. □

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Corollary 2.0.1. *If V is finitely spanned, and $\mathfrak{B} = \{v_1, \dots, v_n\}$ is linearly independent, then \mathfrak{B} can be extended to a basis for V .*

i.e. $\exists w_1, \dots, w_r \in V$, s.t. $\{v_1, \dots, v_n, w_1, \dots, w_r\}$ is a basis for V .

Proof. Let $m = \dim V$, so $n \leq m$. (By theorem).

case 1:

\mathfrak{B} is already a basis ($n = m$). done

Case 2: \mathfrak{B} is not a basis, so $\text{span } \mathfrak{B} \neq V$, so $\exists w_1 \in V \setminus \mathfrak{B}$. □

Theorem 2.0.1. *For any V.S. V , if $\mathfrak{B} \subseteq V$ is linearly independent, then \mathfrak{B} can be extended to a basis for V . [use axiom of choice]*

Example: Let $\mathfrak{B} = \{\cos(nx), n \geq 0\} \cup \{\sin(nx) : n > 0\} \cup \{e^x\}$.

This \mathfrak{B} can be extended to a basis \mathfrak{B}' for $\mathbb{R}^{[0,1]}$.

$$|\mathfrak{B}'| = 2^{\aleph_0}$$

Recall: If $\{v_1, \dots, v_n\} \subseteq V$ is linearly independent. Say $\{v_1, \dots, v_n\}$ is a maximal linearly independent set, if $\forall w \in V \setminus \{v_1, \dots, v_n\}$, $\{v_1, \dots, v_n, w\}$ is linearly dependent.

Corollary 2.0.2. *If V is a finitely spanned set, then every basis is a maximal linearly independent set, and vice versa.*

More generally,

Definition 2.0.1. *Let V be a V.S., a subset $\mathfrak{B} \subseteq V$ is a **maximal linearly independent set** if*

- \mathfrak{B} is linearly independent
- $\forall w \in V \setminus \mathfrak{B}$, $\mathfrak{B} \cup \{w\}$ is linearly dependent.

Theorem 2.0.2. *In any V.S. V , every basis is a maximal linearly independent set, and vice versa.*

Definition 2.0.2. *A **mininal spanning set** is a set \mathfrak{B} such that*

- $\text{span } \mathfrak{B} = V$

- $\forall w \in \mathfrak{B}, \text{span}(\mathfrak{B} \setminus \{w\}) \neq V$

Theorem 2.0.3. *In every vector space V ,*

1. *Every basis is a minimal spanning set and vice versa*
2. *Every spanning set can be "shrunk" to a basis*
i.e. if $\text{span } \mathfrak{B} = V$, then $\exists \mathfrak{B}' \subseteq \mathfrak{B}$ s.t. \mathfrak{B}' is a basis for V .

Proof. For (2), already proved when \mathfrak{B} is countable. Can extend the proof to uncountable "well-ordering \mathfrak{B} ".

To find a basis for $\mathbb{R}^{[0,1]}$

1. start with $\mathfrak{B} = \mathbb{R}^{[0,1]}$
2. well-order \mathfrak{B} ("enumerates" \mathfrak{B})
3. use the enumeration to shrink \mathfrak{B} to a basis

□

2.1 Jan 24

Review: \mathbb{Z}_n = the set of the congruence classes, $x \equiv y \pmod{m} \iff m \mid x - y$

Revisit: $[0] = \{qm : a \in \mathbb{Z}\} = m\mathbb{Z}$.

$-m\mathbb{Z}$ is collapsed to become zero

$-x \equiv y \pmod{n} \iff x = y \in m\mathbb{Z}$.

-advanced notation: $\mathbb{F}/m\mathbb{Z}$.

Version of this:

- $(\mathbb{Z}, +, \cdot) \rightarrow$ a vector space V .
- $(m\mathbb{Z}) \rightarrow$ a subspace of V .

Definition 2.1.1. *Fix a V.S. V over \mathbb{F} , and a subspace W .*

For $x, y \in V$ say $x \equiv y \pmod{W}$, if $x - y \in W$.

Claim: $\equiv \pmod{W}$ is an equiv relation on V .

Proof. For transitivity:

Assume $x, y, z \in V$, $x \equiv y \pmod{W}$ and $y \equiv z \pmod{W}$, by definition, $x - y \in W$, $y - z \in W$.

Then $x - z = (x - y) + (y - z) \in W$ since W is closed under addition.

Then by definition, $x \equiv z \pmod{W}$.

□

Notation 2.1.1. *Define V, W as before:*

For $x \in V$,

$$x + W := \{x + w : w \in W\}$$

(x is fixed, add x to every vector on W). $x + W$ is called **translation of W by x** , or **coset of W through x** .

Claim: V, W as before, for any $x \in V$, the equivalence class (congruence class) of $\equiv \pmod{W}$ containing x is $x + W$.

if $y \equiv x \pmod{W}$, and $w \in W$, then $y \equiv x + w \pmod{W}$.

Proof. For any $y \in V$, $y \in$ the equiv of $\equiv \pmod{W}$ containing x

$$\iff y \equiv x \pmod{W}$$

$$\iff y - x \in W$$

$$\iff y - x = w, \text{ for some } w \in W$$

$$\iff y = x + w$$

$$\iff y \in x + W$$

□

Remark: For $x \in V$, the span class of $\equiv \pmod{W}$ containing x is

$$\{y \in V, y \equiv x \pmod{W}\}$$

Now define

$$\begin{aligned} V/W &:= \text{the set of all equiv classes of the } \equiv \pmod{W} \text{ relation} \\ &:= \text{the set of all translations of } W \\ &:= \{x + W : x \in V\} \neq V \end{aligned}$$

Next, we turn V/W into a vector space over \mathbb{F} ,

$$(x + W) \oplus (y + W) := (x + y) + W$$

$$c(x + W) := (cx) + W$$

Issue: Are the operations well-defined? Yes

E.g. check scalar multiplication:

assume $x + W = x_1 + W$, $x \equiv x_1 \pmod{W} \iff x - x_1 \in W$.

need to know: $\forall c \in \mathbb{F}$,

$$\begin{aligned} (cx + W) &= (cx_1) + W \\ \Updownarrow & \quad \quad \quad cx \equiv cx_1 \pmod{W} \\ \Updownarrow & \quad \quad \quad (cx) - (cx_1) \in W \\ & \quad \quad \quad c(x - x_1) \in W \end{aligned}$$