

§1.3 Chapter 1 Notes And Definitions

Definitions:

Subspaces: A subset W of a vector space V over a field is a subspace of V if W is a vector space of F w/ operations of addition & scalar multiplication defined on V .

- i) $x+y \in W \quad \forall x, y \in W$ closed under addition
- ii) $cx \in W$ for $c \in F, x \in W$ closed under multiplication
- iii) $0 \in W$
- iv) $\forall x \in W \exists y \in W$ s.t. $x+y=0$

Theorem 1.3 Let V be a vector space and W a subset of V . Then W is a subspace of V IFF \forall F.A.T for operations on V

- i) $0 \in W$
- ii) $x+y \in W \quad \forall x, y \in W$
- iii) $cx \in W \quad \forall c \in F, x \in W$

Proof: $W \subseteq V$; and $V(V, F)$ is a vector space

\Rightarrow Let W be a subspace of V . Then W is closed under addition and multiplication

Let $0 \in W$ and $0' \in V$ so $x+0 \in W$ and $x+0' \in W \Rightarrow x+0 = x+0' = 0 = 0'$ by (uniquely)

\Leftarrow assume i), ii), iii) hold; WNTST W is a vector space defined on V

WNTST additive inverse in W

$cx \in W \Rightarrow cx + x \in W$ by ii. Let $c = -1 \in \mathbb{R} \Rightarrow -x + x = 0 \in W$

Definition

transpose: $A_{ij}^t = A_{ji}$

Symmetric Matrix: $A^t = A$; for square Matrices only

i) Symmetric Matrices are closed under Addition & Multiplication

Example 1: $P_n(F)$ be all polynomials in $F(F)$ having degree less than or equal to n

i) $P_0(F)$ has degree $-1 \Rightarrow P_0(F) \neq P_n(F)$

ii) $P_x(F) + P_y(F)$ s.t. $\deg(x) \leq n$ $\deg(y) \leq n \Rightarrow P_x(F) + P_y(F) \in P_n(F)$

iii) $\subset P_n(F) \neq P_n(F)$

$\Rightarrow P_n(F) \leq P(F)$

Example 2

Let $C(\mathbb{R})$ be set of all cont's fns; $\mathcal{F}(\mathbb{R}, \mathbb{R})$ set of all fns $\mathbb{R} \rightarrow \mathbb{R}$

$C(\mathbb{R}) \leq \mathcal{F}(\mathbb{R}, \mathbb{R})$

$f(t) = 0$ is the zero function

$f(t) = 0$ is cont's $\in C(\mathbb{R})$

Let $f, g \in C(\mathbb{R}) \Rightarrow (f+g)(t)$ is cont's $\in C(\mathbb{R})$

$f \in C(\mathbb{R}), c \in \mathbb{R} \Rightarrow cf(t) \in C(\mathbb{R}) \therefore C(\mathbb{R}) \leq \mathcal{F}(\mathbb{R}, \mathbb{R})$

Example 3 Let $M_{n \times n}(F)$ be the vector space of all $n \times n$ Matrices over F

Let $D_{n \times n}(F) \subseteq M_{n \times n}(F)$

$D_{n \times n}(F)$ s.t. $\forall i, j$ $D_{ij} = 0$ for $i \neq j \Rightarrow D(0) \in D_{n \times n}(F)$

$A, B \in D_{n \times n}(F) \Rightarrow A+B \in D_{n \times n}(F)$ $A_{ij} + B_{ij} = 0 + 0 = 0$ $i \neq j$

$A \in D_{n \times n}(F), c \in F \Rightarrow cA \in M_{n \times n}(F)$ where $cA_{ij} = 0$ for $i \neq j \Rightarrow cA \in D_{n \times n}(F)$ subpace.

Theorem 1.4

Any intersection of subspaces of V is a subspace of V .

Proof: Let $Y, W \leq V$ consider $Y \cap W$

$$i) \quad 0 \in Y \text{ and } 0 \in W \Rightarrow 0 \in Y \cap W$$

$$ii) \quad \text{Let } \forall x, z \in Y, \forall x, z \in W \Rightarrow x, z \in Y \cap W$$

$$\text{If } \forall x, z \in Y \text{ and } x, z \in W \Rightarrow x, z \in Y \cap W$$

If $Y \cap W = \emptyset$ then $\{0\} \in Y \cap W$ is a subspace by definition

$$\text{Assume } \exists x, z \in Y, \text{ and } W \Rightarrow x, z \in Y \cap W$$

$$\text{Since } Y, W \text{ are subspaces} \Rightarrow x+z \in Y \text{ and } x+z \in W \Rightarrow x+z \in Y \cap W$$

$$\text{Similarly for } c \in Y, c \in W \Rightarrow c \in Y \cap W$$

If Y, W are disjoint then these above hold for all possible values of vectors

including 0

§1.4 Linear Combinations

Linear Combinations: Let V be a vector space and let S be a nonempty subset

If there exists a finite number of vectors $u_1, \dots, u_n \in S$ and finite number of scalars in F a_1, \dots, a_n s.t. $V = a_1 u_1 + \dots + a_n u_n$ we say V is a linear combination of vectors in S .

Linear Systems: determining if a system can be written as a linear combination

$$v = (2, 6, 8) \quad u_1 = (1, 2, 1), \quad u_2 = (-2, -4, -2) \quad u_3 = (0, 2, 3), \quad u_4 = (2, 0, -3) \quad u_5 = (-3, 0, 16)$$

$$a_1 u_1 + a_2 u_2 + a_3 u_3 + a_4 u_4 + a_5 u_5 = v$$

$$a_1(1, 2, 1) + a_2(-2, -4, -2) + a_3(0, 2, 3) + a_4(2, 0, -3) + a_5(-3, 0, 16) = (2, 6, 8)$$

$$1a_1 - 2a_2 + 2a_4 - 3a_5 = 2$$

$$1a_1 - 2a_2 + 2a_4 - 3a_5 = 2$$

$$1a_1 - 2a_2 + 2a_4 - 3a_5 = 2$$

$$2a_1 - 4a_2 + 2a_3 + 0a_4 + 0a_5 = 6 \rightarrow$$

$$0a_1 + 0a_2 + 2a_3 - 4a_4 + 14a_5 = 2 \Rightarrow 1a_3 - 2a_4 + 7a_5 = 1$$

$$1a_1 - 2a_2 + 3a_3 + 3a_4 + 16a_5 = 8$$

$$0a_1 + 0a_2 + 3a_3 - 5a_4 + 19a_5 = 6$$

$$3a_3 - 5a_4 + 19a_5 = 6$$

$$1a_1 - 2a_2 + 2a_4 - 3a_5 = 2$$

$$1a_1 - 2a_2 + 0a_4 + a_5 = -4 \quad a_1 = -4 + 2a_2 - a_5$$

$$1a_3 - 2a_4 + 7a_5 = 1$$

$$1a_3 + 3a_5 = 7$$

$$\Rightarrow a_3 = 7 - 3a_5$$

$$+ 1a_1 - 2a_5 = 3$$

$$\rightarrow 1a_4 - 2a_5 = 3$$

$$a_4 = 3 + 2a_5$$

$$(a_1, a_2, a_3, a_4, a_5) =$$

Defn: Row Echelon Form

- i) first coefficient is a one
- ii) if an unknown coefficient is the first entry in the equation, it is zero in all others
- iii) If a coefficient is one for a given index, it is zero for all others preceding it.

SPAN: Let S be a nonempty subset of a vector space V , the $\text{span}(S)$ is the set consisting of all linear combinations of the vectors in S .

Theorem 1.5: The span of any subset of S of a vector space V is a subspace of V . Any subspace of V that contains S must also contain the SPAN of S .

Proof:

to prove a subset: Just Prove that an arbitrary element in A is in $B \Rightarrow A \subseteq B$

Generates (or spans): A subset S of a vector space V generates or SPANS V if $\text{SPAN}(S) = V$. Then the vectors in S generate V .

§1.5 Linear Algebra

definition: A subset S of a vector space is **linear dependent** if \exists a finite number of distinct vectors $u_1, \dots, u_n \in S$ & scalars a_1, \dots, a_n not all zeroes s.t. $a_1 u_1 + \dots + a_n u_n = 0$, must be non-empty

defn: **trivial Representation**: If $a_1 u_1 + a_2 u_2 + \dots + a_n u_n = 0$ if $a_1 = \dots = a_n = 0 \quad \forall a_i$
this is the trivial representation of 0 , as a linear combination of $\{u_1, \dots, u_n\}$
note: any subset of a vector space containing 0 is a linearly dependent
since \exists nontrivial representation $0 = 1 \cdot 0$ (this leads to N.V. space)

defn: **Linear Independent**: If only the trivial representation \exists , then LI

i) the empty set is LI

ii) A set $\{u\}$ of single non-zero vector is LI

theorem 1.6: Let V be a vector space $\frac{1}{2} S_1 \subseteq S_2 \subseteq V$. If S_1 is linearly dependent then S_2 is LD

corollary: If S_2 is linearly Independent then S_1 is LI

defn: **Minimal Generating Spanning Sets**

finds a basis which generates the $\text{SPAN}(S)$

for if a proper subset of S generates the $\text{SPAN}(S)$, then S is LD.

If no proper subset of S generates the $\text{SPAN}(S)$, then S is LI

theorem 1.7: Let S be a LI subset of a vector space V , and let $v \in V$ s.t. $v \notin S$. then $S \cup \{v\}$ is LD iff $v \in \text{SPAN}(S)$.

\Rightarrow if $S \cup \{v\}$ is linear dependent then $\exists u_1, \dots, u_n$ and

a_1, \dots, a_n not all zero s.t. $a_1 u_1 + \dots + a_n u_n + a_{n+1} v = 0$

$$\Rightarrow v = (a_1 u_1 + \dots + a_n u_n) a_{n+1}^{-1}$$

$$v \in \text{SPAN}(S)$$

\Leftarrow Let $v \in \text{SPAN}(S)$ where $\exists u_1, \dots, u_n \in S$ s.t. $v = a_1 u_1 + \dots + a_n u_n$ $u_i \in S$

where $a_1 u_1 + \dots + a_n u_n = 0$ iff $\forall i, a_i = 0$

$$\Rightarrow a_1 u_1 + \dots + a_n u_n + (-1)v = 0 \quad \forall i=1:n \quad a_i = 0$$

$$= 0 + (-1)v = 0$$

where $v \neq 0 \quad \Rightarrow \{v\}$ is LD $\Rightarrow S \cup \{v\}$ is LD

§1.6 Bases and Dimension

Defn: A basis \mathcal{B} for V vector space, is a linearly independent subset that generates V .
basis does not need to BE finite.

theorem 1.8: Let V be a vector space and $\mathcal{B} = \{u_1, u_2, \dots, u_n\}$ be a subset of V . Then \mathcal{B} is a basis for V iff $\forall v \in V$ can be uniquely expressed as a linear combination of vectors of \mathcal{B} ; that is
 $\forall v \in V \exists a_1, \dots, a_n \in F$ s.t. $v = a_1 u_1 + \dots + a_n u_n$; where a_1, \dots, a_n are unique.

Proof: Let $\mathcal{B} = \{u_1, \dots, u_n\}$ be a basis for V vector space then $\text{SPAN}(\mathcal{B}) = V$

$$\text{then } v \in \text{SPAN}(\mathcal{B}) \Rightarrow v = a_1 u_1 + \dots + a_n u_n \quad \text{and } v = b_1 u_1 + \dots + b_n u_n$$

$$\text{WNTST } (a_1, \dots, a_n) = (b_1, \dots, b_n)$$

$$\Rightarrow 0 = (a_1 - b_1)u_1 + \dots + (a_n - b_n)u_n. \text{ then } a_i - b_i = 0 \forall i \text{ since } \mathcal{B} \text{ is a basis}$$

$$\Rightarrow a_i = b_i \forall i$$

\Leftarrow Let $v \in V$ can be expressed uniquely as a linear comb of vectors in $\mathcal{B} = \{u_1, \dots, u_n\}$

WNTST \mathcal{B} is a basis (LI, generates)

then $\forall v \in V \exists$ unique scalars (a_1, \dots, a_n) s.t. $v = a_1 u_1 + \dots + a_n u_n$

then by Hypothesis $V \subseteq \text{SPAN}(\mathcal{B})$.

conversely each n -tuple (a_1, \dots, a_n) determines a ^{unique} linear combination expressed

$$\text{as } v \in V \quad \text{SPAN}(\mathcal{B}) \subseteq V \quad \Rightarrow \text{SPAN}(\mathcal{B}) = V.$$

Theorem 1.9 If a vector space V is generated by a finite set of S

then some subset of S is a basis for V . V then has a finite basis.

"Any Spanning Subset for a Vector Space can be reduced to a basis for V ."

Theorem 1.10: Let V be a vector space that is generated by a set G w/ exactly n vectors, and let L be a LI subset of V w/ m vectors. then $m \leq n$ and \exists a subset $H \subseteq G$ containing exactly $n-m$ vectors s.t.

$G = L \cup H$ generates V :

Proof: By induction let $m=0$, s.t. $L=\emptyset$. $\Rightarrow H=G$.

now By induction Hypothesis Assume it is true for some $m \in \mathbb{N}$.

WNTST $m+1$ is true.

Let $L = \{v_1, \dots, v_{m+1}\}$ be LI by Corollary **1.6** $\{v_1, \dots, v_m\}$ is LI

by Hypothesis $m \leq n$ and $\exists U \subseteq G$ s.t. $\{u_1, \dots, u_{n-m}\}$ ^{not necessarily LI} s.t. $\text{SPAN}(L \cup H) = V$

By Hypothesis $\{v_1, \dots, v_m\} \cup \{u_1, \dots, u_{n-m}\}$ generates V .

If $\{v_{m+1}\} \in \text{SPAN}(L \setminus \{v_{m+1}\})$ then $L \cup \{v_{m+1}\}$ is LD by **1.7** contradicting

the Hypothesis. where $n-m > 0$. Since $v_{m+1} \notin \text{SPAN}(L \setminus \{v_{m+1}\})$ and $v_{m+1} \in \text{SPAN}(L)$

then $n > m \Rightarrow n \geq m+1$

$\Rightarrow \exists a_1, \dots, a_m, b_1, \dots, b_{n-m}$ s.t. $v_{m+1} = a_1 v_1 + \dots + a_m v_m + b_1 u_1 + \dots + b_{n-m} u_{n-m}$ by Hypothesis Assume

$u_1 = (-b_1^{-1} a_1) v_1 + (-b_1^{-1} a_2) v_2 + \dots + (-b_1^{-1}) v_{m+1} + (-b_1^{-1} b_2) u_2 + \dots + (-b_1^{-1} b_{n-m}) u_{n-m}$

$H = \{u_2, \dots, u_{n-m}\}$. then $u_1 \in \text{SPAN}(L \cup H) \Rightarrow \{v_1, \dots, v_m, u_1, \dots, u_{n-m}\} \subseteq \text{SPAN}(L \cup H)$ ^{By adding $u_1 \in L \cup H$}

Since $L \cup H$ generates V ; by **1.5** since $L \cup H \subseteq V \Rightarrow \text{SPAN}(L \cup H) \subseteq V$

$\Rightarrow \text{SPAN}(L \cup H) = V$ ~~where~~ $\text{SPAN}(L \cup H) \subseteq \text{SPAN}(L \cup H) \Rightarrow V \subseteq \text{SPAN}(L \cup H)$.

by Hypothesis $\{v_1, \dots, v_m, u_1, \dots, u_{n-m}\}$ generates V

$$\Rightarrow \text{SPAN}(L \cup H) = \text{SPAN}(\{v_1, \dots, v_{m+1}, u_1, \dots, u_{n-m}\}) = V$$


then $H \subseteq G$

where $|H| = n - m - 1 = n - (m+1)$. ↙ by construction

Corollary 1: Let V be a vector space w/ a finite basis. Then every basis for V contains the same number of vectors.

Note: there can be many basis which generate the space.

Let β be a finite basis w/ n vectors. Let γ have more than n vectors, s.t. γ is a basis.

 $\{v_1, \dots, v_n\} = \beta$ $\text{SPAN}(\beta) = V$ ↑
generating set
 $\{v_1, \dots, v_{n+1}\} = \gamma$ is LI let $|\gamma| > n$

$\exists S \subseteq \gamma$ s.t. $S = \{v_1, \dots, v_{n+1}\} \subseteq \gamma$

S is LI, and β generates V | β | = n

put $|S| = n+1 > n$ contradiction. ↙ RT then γ is finite and $|\gamma| \leq n = |\beta|$

then reversing the argument yields $|\beta| \leq |\gamma| \Rightarrow |\beta| = |\gamma|$

Finite Dimensional: A vector space is finite dimensional if it has a basis consisting of a finite number of vectors. The unique number of vectors in each basis is called dimension of V denoted $\dim(V)$.

A vector space not finite dimensional is infinite-dimensional.

Definition: A subset $S \subseteq V$ generates (or SPANS V) if $\text{SPAN}(S) = V$

in this case the vectors of S generate or SPAN V .

$\text{SPAN}(S)$ is not necessarily LI; you can have LD vectors in a set which has $\text{SPAN}(S) = V$.

key concepts: leading a generating set to a basis.

Corollary 2: i) Let V have dimension n . Any generating set must have at least n vectors. A generating set w/ exactly n vectors is a basis.

ii) Any LI set w/ exactly n vectors is a basis.

iii) Every LI subset of V can be extended to a basis for V .

Theorem 1.1: Let W be a subspace of a finite dimensional vector space V . then W is finite dimensional s.t. $\dim(W) \leq \dim(V)$.
if $\dim(W) = \dim(V) \Rightarrow W = V$.

Corollary: If W is a subspace of V of a finite dimensional vector space V , then any basis of W can be extended to a basis for V by Corollary 2 of Replacement theorem.

The Lagrange Interpolation Formula: the polynomials $f_0(x), f_1(x), \dots, f_n(x)$

can be defined as:
$$f_i(x) = \frac{(x-c_0)(x-c_1)\dots(x-c_{i-1})(x-c_{i+1})\dots(x-c_n)}{(c_i-c_0)\dots(c_i-c_{i-1})(c_i-c_{i+1})\dots(c_i-c_n)}$$

$$= \prod_{k \neq i}^n \frac{x-c_k}{c_i-c_k}$$

n = degree of polynomial desired construction

these are defined as Lagrange Polynomials, for each $f_i(x)$ of degree n is in $P_n(F)$ by regarding $f_i(x)$ as a polynomial function

$$f_i(x) : F \rightarrow F$$

$$f_i(c_j) = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$$

the property of LP is that $\{f_0, f_1, \dots, f_n\}$ is a linearly Ind. subset of $P_n(F)$

suppose that the zero function is defined as this suppose zero property:

$$\sum_{i=0}^n a_i f_i = 0 \quad \forall a_i \text{ s.t. } (a_0, \dots, a_n)$$

$$\text{Then } \sum_{i=0}^n a_i f_i(c_j) = 0 \text{ for } j=0, \dots, n$$

$$\text{by definition of LP } \sum_{i=0}^n a_i f_i(c_j) = a_j \text{ for } i=j$$

where $a_j = 0 \Rightarrow \{f_0, \dots, f_n\}$ is linearly Ind. w/ $n+1$ vectors s.t. $\dim(P_n(F))$

$= n+1 \Rightarrow \mathcal{B}$ is basis.

since \mathcal{B} is a basis of polynomial function $g \in P_n(F)$ can be expressed in terms of \mathcal{B} as LC.

$$\text{Suppose } g = \sum_{i=0}^n b_i f_i$$

$$\text{then } g(c_j) = \sum_{i=0}^n b_i f_i(c_j) = b_j \text{ for } i=j \text{ by def'n of LP}$$

$$\Rightarrow g = \sum_{i=0}^n g(c_i) f_i$$

here for b_0, \dots, b_n are $n+1$ scalars not necessarily distinct;

then $g(c_j) = b_j$ is the unique representation by LI of \mathcal{B}

this shows the unique polynomial w/ degree $(f) \leq n$ that has unique values at b_j at points c_j in domain ($j=0, 1, \dots, n$)

Note: if $f \in P_n(F)$ & $f(c_i) = 0$ for $n+1$ distinct scalars (c_0, \dots, c_n) in F

then f is the zero function: AKA (the kernel)

§1.7 Maximal Linearly Independent Subsets.

$$n=1 \quad \{f_0(x), f_1(x)\} \quad \{c_0, c_1\}$$

$$f_0(x) = \prod_{\substack{k=0 \\ k \neq 0}}^1 \frac{x-c_k}{c_0-c_k} = \frac{x-c_1}{c_0-c_1}$$

$$f_1(x) = \prod_{\substack{k=0 \\ k \neq 1}}^1 \frac{x-c_k}{c_1-c_k} = \frac{x-c_0}{c_1-c_0}$$

$$f_0(c_0) = 1 \quad f_0(c_1) = \frac{c_1-c_1}{c_0-c_1} = 0$$

$$f_1(c_0) = 0 \quad f_1(c_1) = 1$$

$$a_0 f_0(c_0) + a_1 f_1(c_0) + a_2 f_0(c_1) + a_3 f_1(c_1) = a_0 f_0(c_1) + a_1 f_1(c_1) = a_0 + a_1$$