

## Normed Vector Spaces

### Vector Spaces

Throughout,  $\mathbb{F} = \mathbb{R}$  or  $\mathbb{C}$ . A **vector space** over  $\mathbb{F}$  is a nonempty set  $V$  equipped with two operations: vector addition and scalar multiplication.

$$\begin{array}{ll} V \times V \xrightarrow{+} V & \\ (v, w) \mapsto v + w & \text{Vector Addition} \\ F \times V \rightarrow V & \\ (\alpha, v) \mapsto \alpha v & \text{Scalar Multiplication} \end{array}$$

The vector space is an Abelian group, where  $u, v, w \in V$  and  $\alpha, \beta \in \mathbb{F}$ , we have:

- (i)  $u + (v + w) = (u + v) + w$
- (ii)  $\exists 0_v \in V$  with  $\forall v \in V, 0_v + v = v + 0_v = v$
- (iii)  $(\forall v \in V)(\exists w \in V)$  with  $v + w = 0_v$
- (iv)  $\forall v, w \in V, v + w = w + v$
- (v)  $\alpha(v + w) = \alpha v + \alpha w, (\alpha + \beta)v = \alpha v + \beta v$
- (vi)  $\alpha(\beta w) = (\alpha\beta)w$
- (vii)  $1 \cdot v = v$

#### Remarks:

- (a)  $0_v$  is unique and known as the zero vector.
- (b) The vector  $w$  in (iii) is unique, and denoted  $-v$ .
- (c)  $0 \cdot v = 0_v$
- (d)  $(-1) \cdot v = -v$
- (e) Property (iv) follows from all the other axioms.
- (f) For  $n \in \mathbb{N}$ ,  $n \cdot v = \underbrace{v + v + \cdots + v}_{n \text{ times}}$

### Subspaces

Let  $V$  be a vector space over  $\mathbb{F}$ . A **subspace** is a nonempty subset  $W \subseteq V$  satisfying the following:

- (i)  $w \in W, \alpha \in \mathbb{F} \rightarrow \alpha w \in W$ .
- (ii)  $w_1, w_2 \in W \Rightarrow w_1 + w_2 \in W$ .

**Remark:**  $0_v$  is always a member of any subspace; a subspace is also a vector space.

#### Proposition: Intersection of Subspaces

If  $\{W_i\}_{i \in I}$  is a family of subspaces of  $V$ , then,  $\bigcap W_i$  is a subspace of  $V$ .

**Proposition: Union of Subspaces**

It is not the case that the union of subspaces of  $V$  is also a subspace. For example, consider  $\mathbb{R}^2$  with the traditional vector space operations:

$$\begin{aligned} \begin{pmatrix} x \\ y \end{pmatrix} + \begin{pmatrix} x' \\ y' \end{pmatrix} &= \begin{pmatrix} x + x' \\ y + y' \end{pmatrix} \\ \alpha \begin{pmatrix} x \\ y \end{pmatrix} &= \begin{pmatrix} \alpha x \\ \alpha y \end{pmatrix} \end{aligned}$$

If  $W_1, W_2 \subseteq V$  are subspaces such that  $W_1 \cup W_2$  is a subspace, then  $W_1 \subseteq W_2$  or  $W_2 \subseteq W_1$ .

**Generated Subspaces**

Let  $S \subseteq V$  be any subset of a vector space  $V$ . Then,

$$\text{span}(S) = \left\{ \sum_{j=1}^n \alpha_j v_j \mid \alpha_1, \dots, \alpha_n \in \mathbb{F}, v_1, \dots, v_n \in S \right\}$$

**Remarks:**

- $\text{span}(S) \subseteq V$  is a subspace.
- $\text{span}(S) = \bigcap W$ , where  $S \subseteq W$  and  $W \subseteq V$  is a subspace. Thus,  $\text{span}(S)$  is the “smallest” subspace containing  $S$ , or the subspace generated by  $S$ .

**Proposition: Quotient Group on Vector Space**

Let  $V$  be a vector space, and let  $W \subseteq V$  is a subspace. Define  $u \sim_W v \leftrightarrow u - v \in W$ .

- (1)  $\sim_W$  is an equivalence relation.
- (2) If  $[v]_W$  denotes the equivalence class of  $v$ , then  $[v]_W = v + W = \{v + w \mid w \in W\}$ .
- (3)  $V/W := \{[v]_W \mid v \in V\}$  is a vector space with  $[v_1]_W + [v_2]_W = [v_1 + v_2]_W$  and  $\alpha[v]_W = [\alpha v]_W$ .

**Proof of (1):**

- Reflexive:  $u \sim_W u$ , since  $u - u = 0 \in W$ .
- Transitive: Suppose  $u \sim_W v$ , and  $v \sim_W z$ . Then,  $u - v \in W$ , and  $v - z \in W$ . So,  $(u - v) + (v - z) \in W$ , so  $u - z \in W$ . Whence,  $u \sim_W z$ .
- Symmetric: If  $u \sim_W v$ , then  $u - v \in W$ , so  $-1 \cdot (u - v) \in W$ , so  $v - u \in W$ . Whence,  $v \sim_W u$ .

**Proof of (2):**

$$\begin{aligned} [v]_W &= \{u \in V \mid u \sim_W v\} \\ &= \{u \in V \mid u - v \in W\} \\ &= \{u \in V \mid u = v + w \text{ some } w \in W\} \\ &= \{v + w \mid w \in W\} \\ &= v + W \end{aligned}$$

**Proof of (3):** Prove that the operation is well-defined.

## Bases

Let  $V$  be a vector space and  $S \subseteq V$  be a subset.

- (1)  $S$  is said to be spanning for  $V$  if  $\text{span}(S) = V$ .
- (2)  $S$  is linearly independent if, for  $\sum_{j=1}^n \alpha_j v_j = 0_v$  with  $\alpha_1, \dots, \alpha_n \in \mathbb{F}$ ,  $v_1, \dots, v_n \in S$ , then  $\alpha_1 = \alpha_2 = \dots = \alpha_n = 0$ .
- (3)  $S$  is a basis for  $V$  if  $S$  is linearly independent and spanning for  $V$ .

### Proposition: Existence of Basis

Every vector space admits a basis. If  $B_0 \subseteq V$  is linearly independent,  $\exists B \subseteq V$  such that  $B$  is a basis and  $B \supseteq B_0$ .

**Background:** A relation on a set  $X$  is a subset  $R \subseteq X \times X$ . If  $R$  is reflexive ( $x \sim x$ ), transitive ( $x \sim y, y \sim z \rightarrow x \sim z$ ), and antisymmetric ( $x \sim y, y \sim x \rightarrow x = y$ ), then  $R$  is an ordering, and we write  $x \leq y$ .

If  $\leq$  is an ordering of  $X$  such that  $\forall x, y \in X$ ,  $x \leq y$  or  $y \leq x$ , then  $\leq$  is a total (or linear) ordering.

Let  $\leq$  be an ordering of  $X$ , let  $Y \subseteq X$ . An upper bound for  $Y$  is an element  $u \in X$  such that  $y \leq u \forall y \in Y$ . A maximal element in  $X$  is an element  $m \in X$  such that  $x \in X$ ,  $x \geq m \rightarrow x = m$ .

**Example:**  $\mathbb{N}$  under the division ordering defines  $a \leq b \Leftrightarrow a|b$ . If we want to find the maximal elements of  $A = \{2, 6, 9, 12\}$ , we would see that they are 9 and 12 (since no element of  $A$  can be divided by 9 and 12). Meanwhile,  $\mathbb{N}$  itself has no maximal elements.

This leads us to ask: given an ordered set,  $(X, \leq)$ , does  $X$  admit maximal elements.

**Zorn's Lemma (or Axiom):** Let  $(X, \leq)$  be an ordered set. Suppose that every totally ordered subset,  $Y \subseteq X$  has an upper bound in  $X$ . Then,  $X$  admits at least one maximal element.

The proof of Zorn's Lemma relies on the Axiom of Choice (and Zorn's Lemma is equivalent to the Axiom of Choice).

**Proof:** Let  $X = \{D \mid B_0 \subseteq D \subseteq V\}$  with  $D$  linearly independent. Since  $B_0 \subseteq X$ ,  $X \neq \emptyset$ . Define  $D, E \in X$ ,  $D \leq E \Leftrightarrow D \subseteq E$ . We will show that  $X$  has a maximal element.

Consider any totally ordered subset,  $Y = \{D_i\}_{i \in I}$ . Consider  $D = \bigcup D_i$ . Clearly,  $B_0 \subseteq D \subseteq V$ . Suppose  $\sum \alpha_k v_k = 0_v$  with  $v_1, \dots, v_n \in D$ . Therefore,  $\exists D_j$  with  $v_1, \dots, v_n \in D_j$  because  $Y$  is totally ordered. However, by definition,  $D_j$  is a linearly independent set — therefore,  $\alpha_k = 0$ . Thus,  $D$  is linearly independent.

Since  $D$  is linearly independent, and  $B_0 \subseteq D$ , it must be the case that  $D \in X$ .  $D$  is also an upper bound for  $Y$ . So, by Zorn's Lemma,  $X$  has a maximal element,  $B$ .

So,  $B_0 \subseteq B \subseteq V$ ,  $B$  is independent, and  $B$  is maximal in  $X$ . We claim that  $B$  is a basis for  $V$ . Suppose toward contradiction that  $\exists v \in V$  such that  $v \notin \text{span}(B)$ . Consider  $B' = B \cup \{v\}$ .

Then,  $B_0 \subseteq B'$ , and  $B'$  is linearly independent — if  $\sum \alpha_k v_k + \alpha v = 0$ , where  $v_1, \dots, v_n \in B$ , then either:

- If  $\alpha = 0$ , then  $\alpha_k v_k = 0 \Rightarrow \alpha_k = 0$ .

- If  $\alpha \neq 0$ , then  $\sum \alpha_k v_k = -\alpha v$ , which means  $v \in \text{span}(B)$ .  $\perp$

Thus, we have a linearly independent set,  $B'$ , with  $B \subseteq B'$ , and  $B_0 \subseteq B'$ . Therefore,  $B' \in X$ . However, this contradicts the maximality of  $B$ . Therefore,  $\text{span}(B) = V$ , and  $B$  is a basis for  $V$ .

## Examples: Vector Spaces

(1)  $n$ -Dimensional Vectors:

$$\mathbb{F}^n = \left\{ \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \mid x_j \in \mathbb{F} \right\}$$

$$\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} + \begin{pmatrix} y_1 \\ \vdots \\ y + n \end{pmatrix} = \begin{pmatrix} x_1 + y + 1 \\ \vdots \\ x_n + y + n \end{pmatrix}$$

$$\alpha \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} \alpha x_1 \\ \vdots \\ \alpha x_n \end{pmatrix}$$

$$B = \{e_1, \dots, e_n\}$$

where  $e_i$  denotes the unit vector at position  $i$ .

(2)  $m \times n$  Matrices:

$$\mathbb{M}_{m,n}(\mathbb{F}) = \left\{ \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix} \mid a_{ij} \in \mathbb{F} \right\}$$

$$(a_{ij}) + (b_{ij}) = (a_{ij} + b_{ij})$$

$$\alpha(a_{ij}) = (\alpha a_{ij})$$

$$B = \{e_{ij}\}$$

where  $e_{ij}$  denotes a matrix of 0 everywhere except column  $i$  and row  $j$ .

(3) Functions with domain  $\Omega$ :

$$\mathcal{F}(\Omega, \mathbb{F}) = \{f \mid f : \Omega \rightarrow \mathbb{F}\}$$

$$(f + g)(x) = f(x) + g(x)$$

$$(\alpha f)(x) = \alpha f(x)$$

(4) Bounded functions with domain  $\Omega$ :

$$\ell_\infty(\Omega, \mathbb{F}) = \{f \in \mathcal{F}(\Omega, \mathbb{F}) \mid \|f\|_\infty \leq \infty\}$$

$$\|f\|_\infty = \sup_{x \in \Omega} |f(x)|$$

Exercises:

- Triangle Inequality:  $\|f + g\|_u \leq \|f\|_u + \|g\|_u$
- Scalar Multiplication/Absolute Homogeneity:  $\|\alpha f\|_u = |\alpha| \|f\|_u$

- Positive Definite:  $\|f\|_u = 0 \Rightarrow f = 0$

**Proof of Triangle Inequality:** Given  $x \in \Omega$ ,

$$\begin{aligned} |(f+g)(x)| &= |f(x) + g(x)| \\ &\leq |f(x)| + |g(x)| \\ &\leq \|f\|_u + \|g\|_u \end{aligned}$$

Therefore,

$$\begin{aligned} \sup |(f+g)(x)| &\leq \|f\|_u + \|g\|_u \\ \|f+g\|_u &\leq \|f\|_u + \|g\|_u \end{aligned}$$

(5) Continuous functions on closed and bounded intervals:

$$C([a, b], \mathbb{F}) = \{f : [a, b] \rightarrow \mathbb{F} \mid f \text{ continuous}\}$$

Check that  $C([a, b], \mathbb{F}) \subseteq \ell_\infty([a, b], \mathbb{F})$  is a subspace.

(6) Let  $f : [a, b] \rightarrow \mathbb{R}$  be any function. Let  $\mathcal{P} : a = x_0 < x_1 < x_2 < \cdots < x_n = b$ .

$$\text{var}(f; \mathcal{P}) := \sum_{k=1}^n |f(x_k) - f(x_{k-1})|$$

$$\text{var}(f) = \sup_{\mathcal{P}} \text{var}(f; \mathcal{P})$$

$$\text{BV}([a, b]) = \{f : [a, b] \rightarrow \mathbb{R} \mid \text{var}(f) < \infty\}$$

$$\|f\|_{\text{BV}} = |f(a)| + \text{var}(f)$$

$\text{BV}([a, b])$  is a vector space.

**Question:** Is  $\mathbb{1}_{\mathbb{Q}} \in \text{BV}([0, 1])$ ?