

ECE 586: Vector Space Methods
Lecture 10 Flip Video: Vector Spaces

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3.3: Vector Spaces

Definition: A **vector space** consists of the following,

- 1 a **field** F of **scalars** (where $0, 1 \in F$ denote the add. and mult. identities)
- 2 a **set** V of **vectors** (which are decorated by an underline in these notes)
- 3 a **binary operation called vector addition**, which maps any pair of vectors $\underline{v}, \underline{w} \in V$ to a vector $\underline{v} + \underline{w} \in V$ satisfying four conditions:
 - 1 vector addition is commutative: $\underline{v} + \underline{w} = \underline{w} + \underline{v}$
 - 2 vector addition is associative: $\underline{u} + (\underline{v} + \underline{w}) = (\underline{u} + \underline{v}) + \underline{w}$
 - 3 there is a unique vector $\underline{0} \in V$ such that $\underline{v} + \underline{0} = \underline{v}, \forall \underline{v} \in V$
 - 4 to each $\underline{v} \in V$ there is a unique vector $-\underline{v} \in V$ such that $\underline{v} + (-\underline{v}) = \underline{0}$
- 4 a **binary operation called scalar multiplication**, which maps any $s \in F$ and $\underline{v} \in V$ to a vector $s\underline{v} \in V$ satisfying four conditions:
 - 1 the identity is multiplicative identity of the field: $1\underline{v} = \underline{v}, \forall \underline{v} \in V$
 - 2 scalar multiplication is associative: $(s_1 s_2)\underline{v} = s_1(s_2\underline{v})$
 - 3 scalar multiplication distributes over vector addition: $s(\underline{v} + \underline{w}) = s\underline{v} + s\underline{w}$
 - 4 scalar addition distributes scalar multiplication: $(s_1 + s_2)\underline{v} = s_1\underline{v} + s_2\underline{v}$.

3.3: Vector Space Examples

Example (Standard vector space for F^n)

Let F be a field, and let $V = F^n$ be the set of n -tuples $\underline{v} = (v_1, \dots, v_n)$. For $\underline{w} = (w_1, \dots, w_n) \in F^n$, the sum of \underline{v} and \underline{w} is defined by

$$\underline{v} + \underline{w} = (v_1 + w_1, \dots, v_n + w_n).$$

The scalar product of $s \in F$ and $\underline{v} \in V$ is defined $s\underline{v} = (sv_1, \dots, sv_n)$.

Example (General vector space of functions)

Let X be a non-empty set and let Y be a vector space over F . Consider the set V of all functions mapping X to Y . The vector addition of two functions $f, g \in V$ is the function $(f + g): X \rightarrow Y$ defined by

$$(f + g)(x) = f(x) + g(x) \quad \forall x \in X,$$

where the RHS is defined by vector addition in Y . The scalar product of $s \in F$ and the function $f \in V$ is the function sf defined by $(sf)(x) = sf(x)$ for all $x \in X$, where the RHS is defined by scalar multiplication in Y .

3.3.1: Subspaces

Definition

Let V be a vector space over F . A **subspace** of V is a subset $W \subset V$ which is itself a vector space over F .

Lemma

A non-empty subset $W \subset V$ is a subspace of V if and only if, for every pair $\underline{w}_1, \underline{w}_2 \in W$ and every scalar $s \in F$, the vector $s\underline{w}_1 + \underline{w}_2 \in W$.

Sketch proof in live session (via inheritance from V).

Example

Let A be an $m \times n$ matrix over F . Then, the subset $V \subseteq F^{n \times 1}$ of vectors satisfying $A\underline{v} = \underline{0}$ forms a subspace.

3.3: Linear Combinations

Definition

A vector $\underline{w} \in V$ is said to be a **linear combination** of the vectors $\underline{v}_1, \dots, \underline{v}_n \in V$ provided that there exist scalars $s_1, \dots, s_n \in F$ such that

$$\underline{w} = \sum_{i=1}^n s_i \underline{v}_i.$$

Definition

Let U be a list (or set) of vectors in V . The **span** of U , denoted $\text{span}(U)$, is defined to be the set of all finite linear combinations of vectors in U .

Example

For a vector space V , the span of any list (or set) of vectors in V forms a subspace.

3.3.2: Linear Dependence and Independence

Definition

Let V be a vector space over F . A list of vectors $\underline{u}_1, \dots, \underline{u}_n \in V$ is called **linearly dependent** if there are scalars $s_1, \dots, s_n \in F$, not all equal to 0, such that

$$\sum_{i=1}^n s_i \underline{u}_i = \underline{0}.$$

A list that is not linearly dependent is called **linearly independent**.

Similarly, a subset $U \subset V$ is called linearly dependent if there is a finite list $\underline{u}_1, \dots, \underline{u}_n \in U$ of distinct vectors that is linearly dependent. Otherwise, it is called linearly independent.

Example

For $V = \mathbb{R}^4$, vectors $\underline{v}_1 = (1, 1, 0, 0)$, $\underline{v}_2 = (0, 1, 1, 0)$, $\underline{v}_3 = (0, 0, 1, 1)$ are linearly independent because $\underline{u} = s_1 \underline{v}_1 + s_2 \underline{v}_2 + s_3 \underline{v}_3 \neq \underline{0}$ unless $s_1 = s_2 = s_3 = 0$ (e.g., $u_1 \neq 0$ if $s_1 \neq 0$, $u_4 \neq 0$ if $s_3 \neq 0$, and $u_2 \neq 0$ if $s_2 \neq 0 \wedge s_1 = 0$).

3.3.2: Basis

Definition

Let V be a vector space over F . Let $\mathcal{B} = \{\underline{v}_\alpha \mid \alpha \in A\}$ be a subset of linearly independent vectors from V such that every $\underline{v} \in V$ can be written as a finite linear combination of vectors from \mathcal{B} . Then, the set \mathcal{B} is a **Hamel basis** for V . If V has a finite basis, it is called **finite-dimensional**.

From this, a basis decomposition $\underline{v} = \sum_{i=1}^n s_i \underline{v}_{\alpha_i}$ must be unique:

The difference between any two distinct decompositions produces a finite linear dependency in the basis and, hence, a contradiction.

Theorem

Every vector space has a Hamel basis.

3.3.2: Standard Basis

Example

Let F be a field and let $U \subset F^n$ be the list of vectors $\underline{e}_1, \dots, \underline{e}_n$ defined by

$$\begin{aligned}\underline{e}_1 &= (1, 0, \dots, 0) \\ \underline{e}_2 &= (0, 1, \dots, 0) \\ &\vdots \\ \underline{e}_n &= (0, 0, \dots, 1).\end{aligned}$$

For any $\underline{v} = (v_1, \dots, v_n) \in F^n$, we have

$$\underline{v} = \sum_{i=1}^n v_i \underline{e}_i. \quad (1)$$

Thus, the collection $U = \{\underline{e}_1, \dots, \underline{e}_n\}$ spans F^n . Since $\underline{v} = \underline{0}$ in (1) if and only if $v_1 = \dots = v_n = 0$, U is linearly independent. Accordingly, the set U is a basis for F^n . This basis is termed the **standard basis** of F^n .

3.3.2: Dimension

Theorem

Let V be a finite-dimensional vector space that is spanned by a finite set of vectors $W = \{\underline{w}_1, \dots, \underline{w}_n\}$. If $U = \{\underline{u}_1, \dots, \underline{u}_m\} \subset V$ is a linearly independent set of vectors, then $m \leq n$.

Proof in live session.

Thus, all bases have the same number of vectors.

Definition

The **dimension** of a finite-dimensional vector space is the number of elements in any basis for V . It is denoted by $\dim(V)$.

3.3.2: Invertibility

Lemma

Let $A \in F^{n \times n}$ be an invertible matrix. Then, the columns of A form a basis for F^n . Similarly, the rows of A will also form a basis for F^n .

Proof in live session.

Theorem

Let A be an $n \times n$ matrix over F whose columns, denoted by $\underline{a}_1, \dots, \underline{a}_n$, form a linearly independent set of vectors in F^n . Then A is invertible.

Proof in live session.

- To continue studying after this video –
 - Try the required reading: Course Notes EF 3.1 - 3.3
 - Or the recommended reading: LADR Ch. 1, Ch. 2
 - Also, look at the problems in Assignment 4