

Lecture 6 - Vector Spaces

COMP1046- Maths for Computer Scientists

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By the end of this lecture we will have learned:

- ⊙ Vector Spaces
- ⊙ Linear Dependence
- ⊙ Linear Span

Based on Sections 8.1 to 8.4 of the textbook (Neri 2018).

Vector Spaces



Definition

- ⊙ Let E to be a non-null set ($E \neq \emptyset$) and \mathbb{K} to be a *scalar set* (typically $\mathbb{K} = \mathbb{R}$).
- ⊙ Let us name *vectors* the elements of the set E .
- ⊙ Let “+” be an internal composition law, $E \times E \rightarrow E$.
- ⊙ Let “.” be an external composition law, $\mathbb{K} \times E \rightarrow E$.

The triple $(E, +, \cdot)$ is said *vector space* of the vector set E over the *scalar field* $(\mathbb{K}, +, \cdot)$ if and only if the ten *vector space axioms* are verified.

continued...

Vector space axioms (1 to 5)

Definition

- ⊙ E is closed with respect to the internal composition law:
 $\forall \mathbf{u}, \mathbf{v} \in E : \mathbf{u} + \mathbf{v} \in E$
- ⊙ E is closed with respect to the external composition law:
 $\forall \mathbf{u} \in E \text{ and } \forall \lambda \in \mathbb{K} : \lambda \mathbf{u} \in E$
- ⊙ commutativity for the internal composition law:
 $\forall \mathbf{u}, \mathbf{v} \in E : \mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$
- ⊙ associativity for the internal composition law:
 $\forall \mathbf{u}, \mathbf{v}, \mathbf{w} \in E : \mathbf{u} + (\mathbf{v} + \mathbf{w}) = (\mathbf{u} + \mathbf{v}) + \mathbf{w}$
- ⊙ neutral element for the internal composition law:
 $\forall \mathbf{u} \in E : \exists ! \mathbf{o} \in E | \mathbf{u} + \mathbf{o} = \mathbf{u}$

where \mathbf{o} is the null vector.

Vector space axioms (6 to 10)

Definition

- ⊙ opposite element for the internal composition law:
 $\forall \mathbf{u} \in E : \exists ! -\mathbf{u} \in E \mid \mathbf{u} + -\mathbf{u} = \mathbf{o}$
- ⊙ associativity for the external composition law: $\forall \mathbf{u} \in E$ and $\forall \lambda, \mu \in \mathbb{K} : \lambda (\mu \mathbf{u}) = (\lambda \mu) \mathbf{u} = \lambda \mu \mathbf{u}$
- ⊙ distributivity 1: $\forall \mathbf{u}, \mathbf{v} \in E$ and $\forall \lambda \in \mathbb{K} : \lambda (\mathbf{u} + \mathbf{v}) = \lambda \mathbf{u} + \lambda \mathbf{v}$
- ⊙ distributivity 2: $\forall \mathbf{u} \in E$ and $\forall \lambda, \mu \in \mathbb{K} : (\lambda + \mu) \mathbf{u} = \lambda \mathbf{u} + \mu \mathbf{u}$
- ⊙ neutral elements for the external composition law:
 $\forall \mathbf{u} \in E : \exists ! 1 \in \mathbb{K} \mid 1 \mathbf{u} = \mathbf{u}$

where \mathbf{o} is the null vector.

Vector Spaces

Example

The set of numeric vectors $E = \mathbb{R}^3$ with scalar set \mathbb{R} , vector sum and scalar product form a vector space.

- ⊙ Example of commutativity:

$$(2, 3, 1) + (0, -1, 2) = (0, -1, 2) + (2, 3, 1) = (2, 2, 3).$$

- ⊙ Example of distributivity 1:

$$2 \times ((1.5, 3, -1.4) + (0, -1.5, 2)) =$$

$$2 \times (1.5, 3, -1.4) + 2 \times (0, -1.5, 2) = (3, 3, 3.2).$$

Example

The set of matrices $\mathbb{R}_{m,n}$, the sum between matrices and the product of a scalar by a matrix, $(\mathbb{R}_{m,n}, +, \cdot)$.

Definition

Let $(E, +, \cdot)$ be a vector space, $U \subset E$, and $U \neq \emptyset$.

The triple $(U, +, \cdot)$ is a *vector subspace* of $(E, +, \cdot)$ if $(U, +, \cdot)$ is a vector space over the same field \mathbb{K} with respect to both the composition laws.

Proposition

Let $(E, +, \cdot)$ be a vector space, $U \subset E$, and $U \neq \emptyset$.

The triple $(U, +, \cdot)$ is a vector subspace of $(E, +, \cdot)$ if and only if U is closed with respect to both the composition laws $+$ and \cdot , i.e.

- ⊙ $\forall \mathbf{u}, \mathbf{v} \in U : \mathbf{u} + \mathbf{v} \in U$
- ⊙ $\forall \lambda \in \mathbb{K} \text{ and } \forall \mathbf{u} \in U : \lambda \mathbf{u} \in U.$

This proposition shows that we do not need to prove all 10 axioms to show $(U, +, \cdot)$ is a vector subspace of another. We just need to prove closure of the two composition laws.

Vector Subspace

Need to prove “if and only if” both ways:

Proof.

Since the elements of U are also elements of E , they are vectors that satisfy the eight axioms regarding internal and external composition laws. If U is also closed with respect to the composition laws then $(U, +, \cdot)$ is a vector space and since $U \subset E$, U is vector subspace of $(E, +, \cdot)$.

If $(U, +, \cdot)$ is a vector subspace of $(E, +, \cdot)$, then it is a vector space. Thus, the ten axioms, including the closure with respect of the composition laws, are valid. □

Example

Let us consider the vector space $(\mathbb{R}^3, +, \cdot)$ and its subset $U \subset \mathbb{R}^3$:

$$U = \{(x, y, z) \in \mathbb{R}^3 \mid 3x + 4y - 5z = 0\}$$

and let us prove that $(U, +, \cdot)$ is a vector subspace of $(\mathbb{R}^3, +, \cdot)$.

We have to prove the closure with respect to the two composition laws.

continued...

Example

1. Let us consider two arbitrary vectors belonging to U , $\mathbf{u}_1 = (x_1, y_1, z_1)$ and $\mathbf{u}_2 = (x_2, y_2, z_2)$. These two vectors are such that

$$3x_1 + 4y_1 - 5z_1 = 0 \text{ and } 3x_2 + 4y_2 - 5z_2 = 0.$$

Let us calculate $\mathbf{u}_1 + \mathbf{u}_2 = (x_1 + x_2, y_1 + y_2, z_1 + z_2)$.

In correspondence to the vector $\mathbf{u}_1 + \mathbf{u}_2$,

$$\begin{aligned} & 3(x_1 + x_2) + 4(y_1 + y_2) - 5(z_1 + z_2) = \\ & = 3x_1 + 4y_1 - 5z_1 + 3x_2 + 4y_2 - 5z_2 = 0 + 0 = 0. \end{aligned}$$

This means that $\forall \mathbf{u}_1, \mathbf{u}_2 \in U : \mathbf{u}_1 + \mathbf{u}_2 \in U$.

continued...

Example

2. Let us consider an arbitrary vector $\mathbf{u} = (x, y, z) \in U$ and an arbitrary scalar $\lambda \in \mathbb{R}$. We know that $3x + 4y - 5z = 0$.
Let us calculate $\lambda\mathbf{u} = (\lambda x, \lambda y, \lambda z)$.

In correspondence to the vector $\lambda\mathbf{u}$,

$$\begin{aligned} 3\lambda x + 4\lambda y - 5\lambda z &= \\ &= \lambda (3x + 4y - 5z) = \lambda 0 = 0. \end{aligned}$$

This means that $\forall \lambda \in \mathbb{R}$ and $\forall \mathbf{u} \in U : \lambda\mathbf{u} \in U$.

Thus, we proved that $(U, +, \cdot)$ is a vector subspace $(\mathbb{R}^3, +, \cdot)$.

Exercise 1

Consider the vector space $(\mathbb{R}^2, +, \cdot)$ and its subsets $V \subset \mathbb{R}^2$ and $W \subset \mathbb{R}^2$:

$$V = \{(x, y) \in \mathbb{R}^2 \mid x + 2y > 1\}$$

$$W = \{(x, y) \in \mathbb{R}^2 \mid 2x = y\}.$$

Show whether or not $(V, +, \cdot)$ and $(W, +, \cdot)$ are vector subspaces of $(\mathbb{R}^2, +, \cdot)$.

Exercise 1: Solution

To be completed...

Linear Dependence

Linear Dependence in n Dimensions

Definition

Let $(E, +, \cdot)$ be a vector space. Let the vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n \in E$ and the scalars $\lambda_1, \lambda_2, \dots, \lambda_n \in \mathbb{K}$.

The *linear combination* of the n vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ by means of n scalars $\lambda_1, \lambda_2, \dots, \lambda_n$ is the vector $\lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \dots + \lambda_n \mathbf{v}_n$.

Definition

Let $(E, +, \cdot)$ be a vector space. Let the vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n \in E$. These vectors are said to be *linearly dependent* if the null vector \mathbf{o} can be expressed as linear combination by means of the scalars $\lambda_1, \lambda_2, \dots, \lambda_n \neq 0, 0, \dots, 0$.

Note: this means at least one $\lambda_i \neq 0$ (not necessarily all).

Linear Dependence in n Dimensions

Definition

Let $(E, +, \cdot)$ be a vector space. Let the vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n \in E$.

These vectors are said to be *linearly independent* if the null vector $\mathbf{0}$ can be expressed as linear combination only by means of the scalars $0, 0, \dots, 0$.

Linear Dependence in 3 Dimensions

Example

Let us consider the following vectors $\in \mathbb{R}^3$

$$\mathbf{v}_1 = (4, 2, 0)$$

$$\mathbf{v}_2 = (1, 1, 1)$$

$$\mathbf{v}_3 = (6, 4, 2).$$

These vectors are linearly dependent since

$$(0, 0, 0) = (4, 2, 0) + 2(1, 1, 1) - (6, 4, 2);$$

that is, \mathbf{v}_3 as a linear combination of \mathbf{v}_1 and \mathbf{v}_2

$$(6, 4, 2) = (4, 2, 0) + 2(1, 1, 1).$$

Linear Dependence in n Dimensions

Theorem

Let $(E, +, \cdot)$ be a vector space. Let the vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n \in E$.

If the n vectors are linearly dependent while $n - 1$ are linearly independent, there is a unique way to express one vector as linear combination of the others:

$$\forall \mathbf{v}_k \in E, \exists! \lambda_1, \lambda_2, \dots, \lambda_{k-1}, \lambda_{k+1}, \dots, \lambda_n \neq 0, 0, \dots, 0 \text{ such that}$$
$$\mathbf{v}_k = \lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \dots + \lambda_{k-1} \mathbf{v}_{k-1} + \lambda_{k+1} \mathbf{v}_{k+1} + \dots + \lambda_n \mathbf{v}_n$$

Note that this means $\lambda_i \neq 0 \quad \forall i = 1, \dots, k-1, k+1, \dots, n$.

Linear Dependence in n Dimensions

Proof.

Let us assume by contradiction that the linear combination is not unique:

⊙ $\exists \lambda_1, \lambda_2, \dots, \lambda_{k-1}, \lambda_{k+1}, \dots, \lambda_n \neq 0, 0, \dots, 0$ such that

$$\mathbf{v}_k = \lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 \dots + \lambda_{k-1} \mathbf{v}_{k-1} + \lambda_{k+1} \mathbf{v}_{k+1} + \dots + \lambda_n \mathbf{v}_n$$

⊙ $\exists \mu_1, \mu_2, \dots, \mu_{k-1}, \mu_{k+1}, \dots, \mu_n \neq 0, 0, \dots, 0$ such that

$$\mathbf{v}_k = \mu_1 \mathbf{v}_1 + \mu_2 \mathbf{v}_2 \dots + \mu_{k-1} \mathbf{v}_{k-1} + \mu_{k+1} \mathbf{v}_{k+1} + \dots + \mu_n \mathbf{v}_n$$

where $\lambda_1, \lambda_2, \dots, \lambda_{k-1}, \lambda_{k+1}, \dots, \lambda_n \neq$

$\mu_1, \mu_2, \dots, \mu_{k-1}, \mu_{k+1}, \dots, \mu_n \neq 0, 0, \dots, 0.$

continued...

Proof.

Under this hypothesis, we can write that

$$\begin{aligned}\mathbf{o} &= (\lambda_1 - \mu_1) \mathbf{v}_1 \\ &+ (\lambda_2 - \mu_2) \mathbf{v}_2 \\ &+ \dots \\ &+ (\lambda_{k-1} - \mu_{k-1}) \mathbf{v}_{k-1} \\ &+ (\lambda_{k+1} - \mu_{k-1}) \mathbf{v}_{k+1} \\ &+ \dots \\ &+ (\lambda_n - \mu_n) \mathbf{v}_n\end{aligned}$$

continued...

Linear Dependence in n Dimensions

Proof.

Since the $n - 1$ vectors are linearly independent

$$\left\{ \begin{array}{l} \lambda_1 - \mu_1 = 0 \\ \lambda_2 - \mu_2 = 0 \\ \dots \\ \lambda_{k-1} - \mu_{k-1} = 0 \\ \lambda_{k+1} - \mu_{k+1} = 0 \\ \dots \\ \lambda_n - \mu_n = 0 \end{array} \right. \Rightarrow \left\{ \begin{array}{l} \lambda_1 = \mu_1 \\ \lambda_2 = \mu_2 \\ \dots \\ \lambda_{k-1} = \mu_{k-1} \\ \lambda_{k+1} = \mu_{k+1} \\ \dots \\ \lambda_n = \mu_n. \end{array} \right.$$

Thus, the linear combination is unique.

□

Linear Dependence in n Dimensions

Proposition

Let $(E, +, \cdot)$ be a vector space and $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ be its n vectors. If one of these vectors is equal to the null vector \mathbf{o} , these vectors are linearly dependent.

Proof.

Let us assume that $\mathbf{v}_n = \mathbf{o}$ and let us pose

$$\mathbf{o} = \lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \dots + \lambda_{n-1} \mathbf{v}_{n-1} + \lambda_n \mathbf{o}.$$

Even if $\lambda_1, \lambda_2, \dots, \lambda_{n-1} = 0, 0, \dots, 0$, the equality is verified for any scalar $\lambda_n \in \mathbb{K}$. Thus, the vectors are linearly dependent. □

Exercise 2

1. Consider the vector space $(\mathbb{R}^3, +, \cdot)$ and the vectors $(1, 0, 2)$, $(2, -1, 1)$, $(3, x, 0)$ from this vector space.
What value(s) of x will make these three vectors linearly dependent?
2. Consider the vector space $(\mathbb{R}^2, +, \cdot)$ and the vectors $(1, 0)$, $(0, 2)$ from this vector space.
Write down a third vector \mathbf{v} such that $(1, 0)$, $(0, 2)$, \mathbf{v} are linearly independent. Otherwise, if it is not possible, explain why.

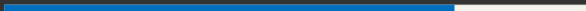
Exercise 2: Solution

To be completed...

Exercise 2: Solution

To be completed...

Linear Span



Definition

Let $(E, +, \cdot)$ be a vector space. The set containing the totality of all the possibly linear combinations of the vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n \in E$ by means of n scalars is named *linear span* (or simply span) and is indicated with $L(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n) \subset E$ or synthetically with L :

$$L(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n) = \{\lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \dots + \lambda_n \mathbf{v}_n \mid \lambda_1, \lambda_2, \dots, \lambda_n \in \mathbb{K}\}.$$

In the case $L(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n) = E$, the vectors are said to span the set E or, equivalently, are said to span the vector space $(E, +, \cdot)$.

Example

The vectors $\mathbf{v}_1 = (1, 0)$, $\mathbf{v}_2 = (0, 2)$, $\mathbf{v}_3 = (1, 1)$ span the entire \mathbb{R}^2 since any point $(x, y) \in \mathbb{R}^2$ can be generated from

$$\lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \lambda_3 \mathbf{v}_3$$

with

$$\lambda_1, \lambda_2, \lambda_3 \in \mathbb{R}.$$

We think of the vectors forming the span as building blocks for the vector space.

Theorem

The span $L(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n)$ with the composition laws is a vector subspace of $(E, +, \cdot)$.

Proof.

In order to prove that $(L, +, \cdot)$ is a vector subspace, using Proposition on Slide 8, it is enough to prove the closure of L with respect to the composition laws. *continued...*

Proof.

1. Let \mathbf{u} and \mathbf{w} be two arbitrary distinct vectors $\in L$. Thus,

$$\mathbf{u} = \lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \dots + \lambda_n \mathbf{v}_n$$

$$\mathbf{w} = \mu_1 \mathbf{v}_1 + \mu_2 \mathbf{v}_2 + \dots + \mu_n \mathbf{v}_n.$$

Let us compute $\mathbf{u} + \mathbf{w}$,

$$\begin{aligned}\mathbf{u} + \mathbf{w} &= \lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \dots + \lambda_n \mathbf{v}_n \\ &\quad + \mu_1 \mathbf{v}_1 + \mu_2 \mathbf{v}_2 + \dots + \mu_n \mathbf{v}_n \\ &= (\lambda_1 + \mu_1) \mathbf{v}_1 + (\lambda_2 + \mu_2) \mathbf{v}_2 + \dots + (\lambda_n + \mu_n) \mathbf{v}_n.\end{aligned}$$

Hence $\mathbf{u} + \mathbf{w} \in L$.

continued...

Proof.

1. Let \mathbf{u} be an arbitrary vector $\in L$ and μ an arbitrary scalar $\in \mathbb{K}$. Thus,

$$\mathbf{u} = \lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \dots + \lambda_n \mathbf{v}_n.$$

Let us compute $\mu \mathbf{u}$:

$$\begin{aligned}\mu \mathbf{u} &= \mu (\lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \dots + \lambda_n \mathbf{v}_n) = \\ &= \mu \lambda_1 \mathbf{v}_1 + \mu \lambda_2 \mathbf{v}_2 + \dots + \mu \lambda_n \mathbf{v}_n.\end{aligned}$$

Hence, $\mu \mathbf{u} \in L$.



Exercise 3

Consider the vector space $(\mathbb{R}^3, +, \cdot)$ and its subset $U \subset \mathbb{R}^3$:

$$U = \{(x, y, z) \in \mathbb{R}^3 \mid 3x + 4y - 5z = 0\}$$

from slide 10.

Show that $(1, 0, \frac{3}{5})$, $(0, 1, \frac{4}{5})$ span U .

Exercise 3: Solution

To be completed...

Summary and next lecture

Summary

- ⊙ Vector Spaces
- ⊙ Linear Dependence
- ⊙ Linear Span

The next lecture

We will learn about the Basis and Dimension of a Vector Space.