

Review of Linear Algebra

Vector Space. = Linear Space.

$$\mathbb{R}^n = \left\{ \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, x_1, \dots, x_n \in \mathbb{R} \right\}$$

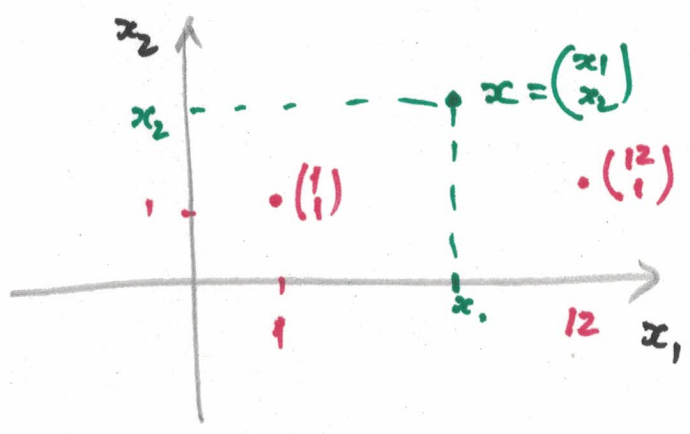
$x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$: n-dimensional vector.

n = dimension of space \mathbb{R}^n

n is a positive integer, $n \geq 1$.

Example.

n=2, \mathbb{R}^2



Addition of vectors.

$$+ : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}^n$$

$$x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, y = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} \longrightarrow x + y = \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \\ \vdots \\ x_n + y_n \end{pmatrix}$$

$(\mathbb{R}^n, +)$ is an abelian group.

$$0 \in \mathbb{R}^n, \text{ 0-vector: } 0 = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$-x = \begin{pmatrix} -x_1 \\ -x_2 \\ \vdots \\ -x_n \end{pmatrix}.$$

Multiplication with scalars ;

Scalar-Vector Multiplication

$$a \in \mathbb{R}, x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n \mapsto a \cdot x = \begin{pmatrix} ax_1 \\ ax_2 \\ \vdots \\ ax_n \end{pmatrix} \in \mathbb{R}^n$$

Convention

$$x \cdot a = a \cdot x = \begin{pmatrix} ax_1 \\ ax_2 \\ \vdots \\ ax_n \end{pmatrix}$$

The set $(\mathbb{R}^n, +, \cdot)$ is called a (real) vector space.
field of scalar. \mathbb{R}
↑
scalar-vector multiplication.

Example 2: $V = \mathbb{C}^n = \left\{ \begin{pmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{pmatrix}, z_1, \dots, z_n \in \mathbb{C} \right\}$

$$z = \begin{pmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{pmatrix}, w = \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix} \mapsto z + w = \begin{pmatrix} z_1 + w_1 \\ z_2 + w_2 \\ \vdots \\ z_n + w_n \end{pmatrix}; a \in \mathbb{C} \rightarrow a \cdot z = \begin{pmatrix} az_1 \\ az_2 \\ \vdots \\ az_n \end{pmatrix}$$

↑
addition
scalar-vector multiplication

$(\mathbb{C}^n, +, \cdot)_{\mathbb{C}}$ is a (complex) vector space.

In general:

$(V, +, \cdot)_{\mathbb{F}}$ is a vector space (over F).
field $(F, +, \cdot) \rightarrow$ For us: $F = \mathbb{R}$
or $F = \mathbb{C}$

set of 1) $(V, +)$: abelian group.
vectors.

2) $(F, +, \cdot)$: field.

[3]. distributivity of \cdot over $+$].

Example: Fix $1 \leq p < \infty$

$$l^p(\mathbb{Z}) = \left\{ x = (x_n)_{n \in \mathbb{Z}} : \sum_{n \in \mathbb{Z}} |x_n|^p < \infty \right\}$$

$$x = (\dots, 0, 0, \underset{\substack{\uparrow \\ \text{0th position}}}{1}, 0, 0, \dots) \in l^p$$

But

$$y = (\dots, \underbrace{1, 1, 1, 1, 1, 1, \dots}_{\text{constant 1}}) \notin l^p$$

If $x = (x_n)_{n \in \mathbb{Z}}$, $y = (y_n)_{n \in \mathbb{Z}}$ are in $l^p(\mathbb{Z})$

then

$x+y = (x_n+y_n)_{n \in \mathbb{Z}}$ is also in $l^p(\mathbb{Z})$.

$a \in \mathbb{C}$, $x = (x_n)_{n \in \mathbb{Z}}$ \rightarrow $a \cdot x = (a \cdot x_n)_{n \in \mathbb{Z}}$

$(l^p(\mathbb{Z}), +, \cdot)$ is a complex vector space.

For $p = \infty$.

$l^\infty(\mathbb{Z}) = \left\{ x = (x_n)_{n \in \mathbb{Z}} : \text{there is } M > 0 \text{ s.t. } |x_n| \leq M, \text{ for every } n \right\}$

Consider the case. $V = \mathbb{R}^n$ or $V = \mathbb{C}^n$.

- Linear Independent vectors, sets of linearly independent vectors
- Spanning sets
- basis

Linear Independence

Definition.

A set of vectors $\{v_1, v_2, \dots, v_m\}$ is said linearly independent

if it satisfies:

If c_1, c_2, \dots, c_m are scalars such that $c_1 v_1 + c_2 v_2 + \dots + c_m v_m = 0$

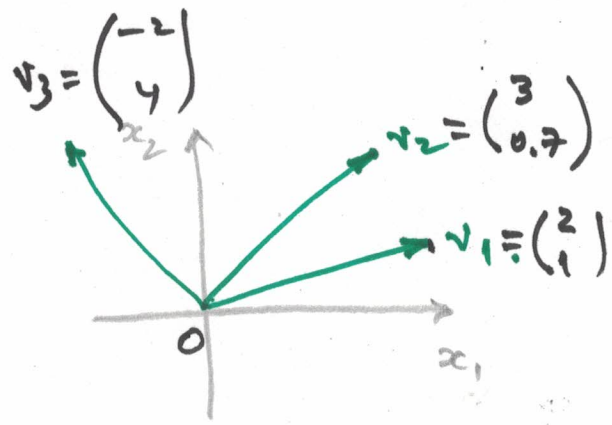
Then $c_1 = c_2 = \dots = c_m = 0$.

In other words:

The only linear combination $c_1 v_1 + c_2 v_2 + \dots + c_m v_m$ that produces the null vector 0 is the combination with all coefficients zero

Examples.

$n=2,$



The set $\{v_1\}$ is linearly independent, as long as $v_1 \neq 0$.

The set $\{v_1, v_2\}$ is linearly independent.

But the set $\{v_1, v_2, v_3\}$ is NOT linearly independent.

Why?: Find $c_2, c_3 \neq 0$ s.t. $v_1 + c_2 v_2 + c_3 v_3 = 0$.

Property IP. $\{v_1, v_2, \dots, v_m\}$ is a set of linearly independent vectors in \mathbb{R}^n (or \mathbb{C}^n) then $m \leq n$.

Let $\{v_1, v_2, \dots, v_m\}$ be a set of vectors in \mathbb{C}^n .

$$V = \begin{pmatrix} | & | & & | \\ v_1 & v_2 & \dots & v_m \\ | & | & & | \end{pmatrix} \in \mathbb{C}^{n \times m}$$

↓
Columns of V are the vectors $v_1 \dots v_m$.

Property. The set $\{v_1 \dots v_m\}$ is linearly independent if and only if the only solution of:

$$V \cdot c = 0, \quad c \in \mathbb{C}^m$$

is the zero vector $c = 0$.

Spanning Sets

Definition A set $\{v_1, v_2, \dots, v_m\} \subset \mathbb{C}^n$ is said spanning for \mathbb{C}^n if:

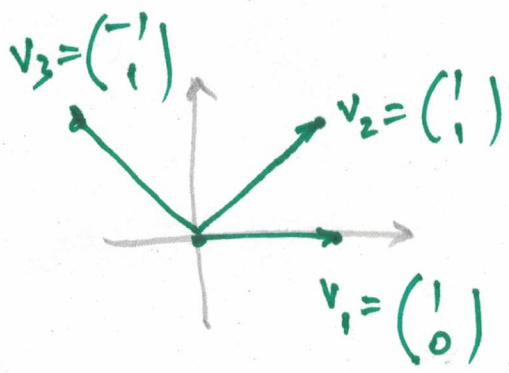
For every $x \in \mathbb{C}^n$ there are $c_1, c_2, \dots, c_m \in \mathbb{C}$ such that $c_1 v_1 + c_2 v_2 + \dots + c_m v_m = x$.

In other words,

Any vector in the space (\mathbb{C}^n) ^{can be} obtained as a linear combination of the spanning set.

Examples

$n=2$



$\{v_1\}$ is NOT a spanning set for \mathbb{C}^2 .

$\{v_1, v_2\}$ is a spanning set for \mathbb{C}^2 .

$\{v_1, v_2, v_3\}$ is a spanning set for \mathbb{C}^2 .

Why: If $x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \rightarrow x = (x_1 - x_2) \begin{pmatrix} 1 \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 1 \\ 1 \end{pmatrix} = c_1 v_1 + c_2 v_2$

where: $c_1 = (x_1 - x_2)$, $c_2 = x_2$.

Property If $\{v_1, v_2, \dots, v_m\}$ is a spanning set for \mathbb{R}^n (or \mathbb{C}^n) then $m \geq n$.

Let $V = (v_1 | v_2 | \dots | v_m) \in \mathbb{C}^{n \times m}$

be the $n \times m$ matrix whose columns are vectors v_1, v_2, \dots, v_m .

Property The set $\{v_1, \dots, v_m\}$ is spanning in \mathbb{C}^n .

if and only if, for any $z \in \mathbb{C}^n$ the linear system:

$$z = V \cdot c$$

has a solution $c \in \mathbb{C}^m$.

Fact. Let $\{v_1, v_2, \dots, v_m\} \subset \mathbb{C}^n$. Let $\{w_1, \dots, w_n\} \subset \mathbb{C}^m$

be the set of vectors in \mathbb{C}^m such that:

$$(v_1 | v_2 | \dots | v_m)^T = (w_1 | \dots | w_n).$$

\leftarrow transpose

Then:

$\{v_1, \dots, v_m\}$ are linearly independent if and only if $\{w_1, w_2, \dots, w_n\}$ is spanning in \mathbb{C}^m .

Definition A set $\{v_1, \dots, v_m\}$ in \mathbb{R}^n (or \mathbb{C}^n) is called a basis if it is simultaneously a linearly independent set and a spanning set.

Property 1. If $\{v_1, \dots, v_m\}$ is a basis in \mathbb{R}^n (or \mathbb{C}^n) then $m = n$.

Property 2. Let $V = (v_1 | \dots | v_m)$ be the matrix whose columns are the vectors $\{v_1, \dots, v_n\}$. Then:

(1) $\{v_1, \dots, v_n\}$ is a basis for \mathbb{R}^n (or \mathbb{C}^n) if and only if.

(2) For any $x \in \mathbb{R}^n$ (or \mathbb{C}^n) the linear system $x = V \cdot c$ has a unique solution $c \in \mathbb{R}^n$ (or \mathbb{C}^n).

Remark. $\{v_1, \dots, v_n\}$ is basis if and only if $\det V \neq 0$.

$\Leftrightarrow V$ is invertible.

Scalar product.

Let V be a vector space.

Definition A function $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{C}$ is called a

scalar product (or inner product) on V if:

(1) (positivity):

(i) For any $x \in V$, $\langle x, x \rangle \geq 0$

(ii) ~~is~~ $\langle x, x \rangle = 0$ if and only if $x = 0$.

(2) (skew-symmetry) For any $x, y \in V$

$\langle x, y \rangle = \overline{\langle y, x \rangle}$ ← complex conjugate

(3) (linearity) For any $x, y, z \in V$, $a, b \in \mathbb{C}$:

$\langle ax + by, z \rangle = a \cdot \langle x, z \rangle + b \langle y, z \rangle.$

Example: If $V = \mathbb{C}^n$

$x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, y = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}, \langle x, y \rangle = x_1 \overline{y_1} + x_2 \overline{y_2} + \dots + x_n \overline{y_n}$