

Linear algebra

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2025

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1 Introduction

In linear algebra we study object, also a point is a linear object. Usually we study this object from an geometry point of view, in this course we see the algebrian point of view. When we do **operations** we are doing algebra. Vector are a kind of linear object, we can see a vector like a point.

In \mathbb{R}^3 we have more kind of linear objects. In \mathbb{R}^4 we start to have problem whit the geometrical point of view. Whit equations i can describe objects in every dimension. Vector spaces is very important in linear algebra. Matrix are very important because we also have functions for transform object.

2 Sets

We denote by \emptyset the empty set.

$$\mathbb{N} = \{0, 1, 2, \dots\} \text{ Natural numbers}$$

$$\mathbb{Z} = \{\dots, -2, -1, 0, 1, 2, \dots\} \text{ Integers}$$

$$\mathbb{Q} = \left\{ \frac{a}{b} : a, b \in \mathbb{Z}, b \neq 0 \right\}$$

\mathbb{R} the set of all real numbers, Every real number is the limit of a sequence of rational numbers.

2.1 Examples

$$A = \{x \in \mathbb{R} : -5 \leq x \leq 2\}$$

$$B = \{x \in \mathbb{Z} : -5 \leq x \leq 2\}$$

$$C = \{x \in \mathbb{R} : x > 1\}$$

2.2 Operation between sets

Given two set A and B we define the following operation:

- Intersection: $A \cap B = \{x : x \in A \cap x \in B\}$
- Union $A \cup B = \{x : x \in A \vee X \in B\}$
- Difference $A \setminus B = \{x : x \in A \wedge x \notin B\}$
- Cartesian product $A \times B = \{(x, y) : x \in A \wedge y \in B\}$

3 Cartesian plane

The cartesian plane is a 2 dimensional object. All the \mathbb{R} can be represented on an «oriented line» where we fixed a special point, namely the origin O. In a similar way we can describe all the elements of $\mathbb{R} \times \mathbb{R} = \mathbb{R}^2 = \{(x, y) : x \in \mathbb{R}, y \in \mathbb{R}\}$ with the cartesian plane. We take

2 oriented lines which are orthogonal and we call the origin O the interaction between these lines (representing the element $(0, 0) \in \mathbb{R}^2$).

The horizontal line is called the axis of abscises or the x-axis. The vertical line is called of ordinate or the y-axis.

Evrey point P describe an element of \mathbb{R}^2 since it is identified by two set number X_P and X_P which are its cordinates, i. e.,

$$\begin{aligned}\text{The y-axis is the following set } &= \{(0, y) \in \mathbb{R}^2\} = \{(0, y) : y \in \mathbb{R}\} \\ &= \{(x, y) \in \mathbb{R}^2 : x = 0 \wedge y \in \mathbb{R}\}\end{aligned}$$

$$\begin{aligned}\text{The x-axis is the following set } &= \{(x, 0) \in \mathbb{R}^2\} = \{(z, 0) : z \in \mathbb{R}\} \\ &= \{(x_1, x_2) \in \mathbb{R}^2 : x_2 = 0\}\end{aligned}$$

Given two point A and $B \in \mathbb{R}^2$, we can evaluate theri distance (euclidean distance) which is the lengh of the segment \overline{AB} :

$$\begin{aligned}d(A, B) &= ||\overline{AB}|| = \sqrt{(x_B - x_A)^2 + (y_B - y_A)^2} \\ &= \sqrt{(x_A - x_B)^2 + (y_A - y_B)^2}\end{aligned}$$

3.1 Vector in the cartesian plane

In general a vector is an arrow. Given a point P in the cartesian plane, i. e., a point of two numbers $(x_P, y_P) \in \mathbb{R}^2$ it also indentified a vector of the plane. The point P identifies the vector $\vec{OP} = (x_P, y_p)$. Sometimes we will also use the notion \vec{v} or v for a vector.

$$v = (v_1, v_2) \in \mathbb{R}^2$$

We have that, from this algebraic point of view, point and vector are equivalent, they are the same object described by a point or set.

3.2 Operations in the cartesian plane

3.2.1 Scalar multiplication

Multiplication by a scalar, i. e., by a real number. Givern a point $A = (x_A, y_A)$ ($\Leftrightarrow Q = (x_A, y_A) = \vec{OA}$) and a scalr $t \in \mathbb{R}$ we define the multiplication by a scalar $tA = (tx_A, ty_A)$.

$$\mathbb{R}^2 \times \mathbb{R} \longrightarrow \mathbb{R}^2$$

$$(A \times t) \longrightarrow tA$$

The effect of the multiplication by a scalar on a vector \vec{a} ub fact the vertor \mathcal{O} is stretched

3.2.2 Sum

Given two points $A = (x_A, y_A)$ ($\Leftrightarrow a = (x_A, y_A)$) and $B = (x_B, y_B)$ the sum is defined as

$$A + B = (x_A + x_B, y_A + y_B)$$

$$\begin{aligned} \mathbb{R}^2 \times \mathbb{R}^2 &\longrightarrow \mathbb{R}^2 \\ (A, B) &\longrightarrow A + B \end{aligned}$$

$a + b$ ($\Leftrightarrow A + B$) is the diagonal of the parallelogram with a and b the following properties hold:

$\forall a, b, c \in \mathbb{R}^2, \forall t \in \mathbb{R}$ we have

- $a + b = b + a$
- $(a + b) + c = a + (b + c)$
- $t(a + b) = ta + tb$
- $a + 0 = a$ when $0 = (0, 0)$ = origin
- $(-1)a = -a$ when $a + (-a) = 0$

3.3 Lines in the cartesian plane

Given a point $A = (x_A, y_A) \in \mathbb{R}^2$, there exist one and only one line ℓ through the origin O and the point A. Now we want to describe the coordinates of all points.

Let us take a generic point $P = (x_P, y_P) \in \ell$ and let us find its coordinate in term of A. The triangles OAH and OPC are similar (they have the same angles) and consequently the side are proportion (by the same factors):

$$|\overline{OK}| = t|\overline{OH}| \quad \text{and} \quad |\overline{PK}| = t|\overline{AH}| \quad \text{for a certain } t \in \mathbb{R}$$

that is

$$\begin{cases} x'_P = tx_A \\ y'_P = ty_A \end{cases} \quad \forall t \in \mathbb{R} \qquad \begin{cases} x = tx_A \\ y = ty_A \end{cases} \quad \forall t \in \mathbb{R}$$

These are the parametric equations of a line through the origin and given a point A.

$$A = (2, 3) \quad \begin{cases} x = 2t \\ y = 3t \end{cases} \quad \forall t \in \mathbb{R}$$

In other words $\ell = \{(x, y) \in \mathbb{R}^2 : x = tx_A, y = ty_A, \forall t \in \mathbb{R}\}$ (1)

The cartesian equation of a line through the origin is

$$y = mx \quad \text{where} \quad m = \frac{y_A}{x_A} \quad \text{slope}$$

$$\ell \left\{ (x, y) \in \mathbb{R}^2 : y = mx, m = \frac{y_A}{x_A} \right\} \quad (2)$$

Let us Considering the set (1) and check that O ans A belong to this set:

$O \in \ell$ because if we take $t = 0$, we get $x = 0 \cdot X_A = 0, y = 0 \cdot y_A = 0$

$$\Rightarrow (0, 0) \in \ell$$

$A \in \ell$ because if we take $t = 1$, we have $x = x_A, y = y_A \Rightarrow A \in \ell$

Let us check that O and A belong to ℓ exploiting the set defined in (2):

$O = (0, 0) \in \ell$ is true, because $0 = m \cdot 0$

$A = (x_A, y_A) \in \ell$ is true, because $y_A = mx_A \Leftrightarrow y_A = \frac{y_A}{x_A} \cdot x_A$

We can check that a point provided by the parametric equations satisfies the cartesian equation.

Let us consider a point (tx_A, ty_A) and check if it satisfies the cartesian equation $y = mx$:

$$ty_A = m t x_A \Leftrightarrow ty_A = \frac{y_A}{x_A} t x_A \Leftrightarrow ty_A = ty_A x$$

Example

Consider the line through the origin and $A = (5, 2)$.

The parametric equation are $\begin{cases} x = 5t \\ y = 2t \end{cases} \quad \forall t \in \mathbb{R}$

$$\begin{aligned} t = 0 & \quad (0, 0), & t = 1 & \quad (5, 2), & t = -2 & \quad (-10, -4), \\ t = \frac{1}{3} & \quad \left(\frac{5}{3}, \frac{2}{3} \right) & \dots & & t = \sqrt{3} & \quad \left(5\sqrt{3}, 2\sqrt{3} \right) \end{aligned}$$

All belong to the line.

The point $(10, 3)$ belong to the line? $t = 2 \quad (10, 4)$

In order to answer to this question we must solve the following linear system

$$\begin{cases} 5t = 10 \\ 2t = 3 \end{cases}$$

where t is the unknown, but this linear system doesn't have solutions!

This $(10, 3)$ doesn't belong to the line The cartesian equation of this line is $y = \frac{2}{5}x \rightarrow 3 = \frac{2}{5} \cdot 10$ is not true! is not an identity

Now, let us describe the parametric equations of a line (not necessary through the origin). Given two points $A = (x_A, y_A)$ and $B = (x_B, y_B)$ we want to describe the coordinates of a

generic point P over the line through A and B . We can consider the line ℓ' parallel to ℓ and passing through the origin. The parametric equations of ℓ' are:

$$\begin{cases} x = t(x_B - x_A) \\ y = t(y_B - y_A) \end{cases} \quad \forall t \in \mathbb{R}$$

The parametric equations of ℓ are

$$\begin{cases} x = x_A + t(x_B - x_A) \\ y = y_A + t(y_B - y_A) \end{cases} \quad \forall t \in \mathbb{R}$$

this

$$\ell \{ (x, y) \in \mathbb{R}^2 : x = x_A + t(x_B - x_A), y = y_A + t(y_B - y_A), \forall t \in \mathbb{R} \}$$

Similarly, the cartesian equation is

$$y = mx + q$$

where $m = \frac{y_B - y_A}{x_B - x_A}$ slope and $q = \frac{y_A x_B - x_A y_B}{x_B - x_A}$ constant term

4 Linear independence (linear dependence)

If two vectors \vec{v}_1 and \vec{v}_2 lie on the same line (through the origin) then they are linearly dependent (from a geometrical point of view).

For an algebraic point of view this means that

$$\vec{v}_1 = t\vec{v}_2 \quad t \in \mathbb{R}$$

Two vectors are linearly independent if and only if they are not linearly dependent.

In general if \vec{w}_1 and \vec{w}_2 are linearly independent vectors than all the vectors $\vec{v} \in \mathbb{R}$ can be written as a linear combination of \vec{w}_1 and \vec{w}_2 , this means that

$$\forall \vec{v} \in \mathbb{R}^2 \quad \vec{v} = a\vec{w}_1 + b\vec{w}_2 \quad \text{for some } a, b \in \mathbb{R}$$

and we say that $\{\vec{w}_1, \vec{w}_2\}$ is a basis of \mathbb{R}^2 .

A special basis is $\{\vec{e}_1, \vec{e}_2\}$ where $\vec{e}_1 = (1, 0)$, $\vec{e}_2 = (0, 1)$ and it is called the canonical basis. $\{\vec{e}_1, \vec{e}_2\}$ is also an orthogonal basis because \vec{e}_1 and \vec{e}_2 are orthogonal vectors.

5 Cartesian space

The cartesian space is $\mathbb{R}^3 = \mathbb{R} \times \mathbb{R} \times \mathbb{R}$ and its elements are triple of real number $(x, y, z) \quad \forall x, y, z \in \mathbb{R}$.

O is a fixed point of the space, called the origin and we also have three orthogonal lines intersecting in the origin.

A point $A \in \mathbb{R}^3$ is identified by three real numbers (x_A, y_A, z_A) where:

x_A is the distance from the plane Oyz

y_a is the distance from the plane Oxz

z_a is the distance from the plane Oxy

Also in the cartesian space, we can compute the Euclidean distance between the origin $\|\overrightarrow{OA}\| = d(O, A)$. Similarly the distance between $A = (x_A, y_A, z_A)$ and $B = (x_B, y_B, z_B)$ is $\|\overrightarrow{AB}\| = d(A, B) = \sqrt{(x_B - x_A)^2 + (y_B - y_A)^2 + (z_B - z_A)^2}$.

Similarly to \mathbb{R}^2 , also in \mathbb{R}^3 the vectors are arrows starting from the origin and they are described by triples of real numbers.

Operations in \mathbb{R}^3

Let us define some operations in \mathbb{R}^3 :

- Scalar multiplication:

$$\begin{aligned}\mathbb{R} \times \mathbb{R}^3 &\longrightarrow \mathbb{R}^3 \\ (t, \vec{v}) &\longmapsto t\vec{v} = (tv_1, tv_2, tv_3) \\ (t, V) &\longmapsto tV = (tv_1, tv_2, tv_3)\end{aligned}$$

where $\vec{v} = V = (v_1, v_2, v_3)$

- Sum (componentwise):

$$\begin{aligned}\mathbb{R}^3 \times \mathbb{R}^3 &\longrightarrow \mathbb{R}^3 \\ (\vec{v}, \vec{w}) &\longmapsto \vec{v} + \vec{w} = (v_1 + w_1, v_2 + w_2, v_3 + w_3) \\ (V, W) &\longmapsto V + W = (v_1 + w_1, v_2 + w_2, v_3 + w_3)\end{aligned}$$

where $\vec{v} = (v_1, v_2, v_3)$ $\vec{w} = (w_1, w_2, w_3)$

Given three vectors in \mathbb{R}^3 , $\vec{u}, \vec{v}, \vec{w}$, we say that they are linearly dependent if there exist three real numbers $a, b, c \in \mathbb{R}$ not all zeros such that

$$a\vec{u} + b\vec{v} + c\vec{w} = 0 = (0, 0, 0)$$

$$\vec{u} = \alpha\vec{v} + \beta\vec{w}$$

$$a\vec{u} + b\vec{v} + c\vec{w} = 0 \Leftrightarrow a\vec{u} = -b\vec{v} - c\vec{w}$$

If the vectors $\vec{u}, \vec{v}, \vec{w}$, are not linearly dependent, then we say that they are linearly independent. In general, given $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$ we say that they are linearly dependent if and only if there exist $a_1, a_2, \dots, a_n \in \mathbb{R}$ not all zeros such that

$$a_1\vec{v}_1 + a_2\vec{v}_2 + \cdots + a_n\vec{v}_n = 0$$

If $\vec{v}_1, \dots, \vec{v}_n$ are not linearly dependent, then we say that they are linearly independent.

In the space, if we have three linearly independent vectors $\vec{w}_1, \vec{w}_2, \vec{w}_3$, then all the vectors $\vec{v} \in \mathbb{R}^3$ can be written as a linear combination of $\vec{w}_1, \vec{w}_2, \vec{w}_3$. And we say the $\{\vec{w}_1, \vec{w}_2, \vec{w}_3\}$ is a basis of \mathbb{R}^3 .

The canonical basis of \mathbb{R}^3 is $\{\vec{e}_1, \vec{e}_2, \vec{e}_3\}$ where $\vec{e}_1 = (1, 0, 0)$, $\vec{e}_2 = (0, 1, 0)$, $\vec{e}_3 = (0, 0, 1)$.

Let us check that $\vec{e}_1, \vec{e}_2, \vec{e}_3$ are linearly independent, we must find if there exist a, b, c not all zeros such that

$$\begin{aligned} a\vec{e}_1 + b\vec{e}_2 + c\vec{e}_3 &= \vec{0} \\ (a, 0, 0) + (0, b, 0) + (0, 0, c) &= \vec{0} \\ (a, b, c) &= (0, 0, 0) \end{aligned}$$

$$\begin{cases} a = 0 \\ b = 0 \\ c = 0 \end{cases} \iff \vec{e}_1, \vec{e}_2, \vec{e}_3 \text{ are l.i.}$$

$\forall \vec{v} \in \mathbb{R}^3$, $\vec{v} = (v_1, v_2, v_3)$, then $\vec{v} = v_1\vec{e}_1 + v_2\vec{e}_2 + v_3\vec{e}_3$.

Example

$$\vec{v} = (5, 3, 2) = 5(1, 0, 0) + 3(0, 1, 0) + 2(0, 0, 1)$$

5.1 Equation of a plane in \mathbb{R}^3

We would like to characterize all the points belonging to some plane in \mathbb{R}^3

$$\begin{array}{ll} \text{The plane} & Oxy = \{(x, y, 0) : x, y \in \mathbb{R}\} \\ & Oyz = \{(0, y, z) : y, z \in \mathbb{R}\} \\ & Oxz = \{(x, 0, z) : x, z \in \mathbb{R}\} \end{array}$$

Given a generic plane in the space, it is uniquely determined by:

- Two intersecting lines
- Three non collinear points
- two linearly independent vectors (identify uniquely a plane through the origin) and a point in \mathbb{R}^3

We start considering a plane described by two linearly independent vectors, i.e., we are describing all the planes through the origin.

We want to describe any point $P = (x_P, y_P, z_P)$ belonging to the plane identified by two linearly independent vectors \vec{v} and \vec{w} .

Remember that two linearly independent vectors can be used for constructing all the vectors belonging to the same plane (think, e.g., to the special case of the cartesian plane). All the points $P = (x_P, y_P, z_P)$ belonging to the plane containing the vectors \vec{v} and \vec{w} are linear combinations of \vec{v} and \vec{w} :

$$P = \overline{OP} = (x_P, y_P, z_P) = s\vec{v} + t\vec{w} \quad \forall s, t \in \mathbb{R}$$

The parametric equations of the plane Π through the origin and containing \vec{v} and \vec{w} are

$$\begin{cases} x = sv_1 + tw_1 \\ y = sv_2 + tw_2 \\ z = sv_3 + tw_3 \end{cases} \quad \forall s, t \in \mathbb{R}$$

In other words

$$\Pi = \{(x, y, z) \in \mathbb{R}^3 : x = sv_1 + tw_1, y = sv_2 + tw_2, z = sv_3 + tw_3, \forall s, t \in \mathbb{R}\}$$

The parametric equations of a plane Π identified by two linearly independent vectors \vec{v} and \vec{w} and through a point $A = (x_A, y_A, z_A)$ are

$$\begin{cases} x = x_A + sv_1 + tw_1 \\ y = y_A + sv_2 + tw_2 \\ z = z_A + sv_3 + tw_3 \end{cases} \quad \forall s, t \in \mathbb{R}$$

$$s = 0, \quad t = 0 \quad \begin{cases} x = x_A \\ y = y_A \\ z = z_A \end{cases}$$

We can obtain the cartesian equation of a plane. If we focus on the equations $\begin{cases} x = x_A + sv_1 + tw_1 \\ y = y_A + sv_2 + tw_2 \end{cases}$ where s and t are unknowns. We can solve this linear system and we obtain the solutions:

$$\begin{cases} s = ax + by + f & a, b, f \in \mathbb{R} \\ t = cx + dy + g & c, d, g \in \mathbb{R} \end{cases}$$

Now we substitute these values of s and t in the third equations:

$$z = z_A + sv_3 + tw_3$$

and we obtain something like $z = m + nx + py$ $m, n, p \in \mathbb{R}$ Cartesian equation of a plane in the space

$$\alpha x + \beta y + \gamma z + \delta = 0$$

Cartesian equation, where $\alpha, \beta, \gamma, \delta \in \mathbb{R}$ depending on $\begin{array}{lll} x_A, & y_A, & z_A \\ v_1, & v_2, & v_3 \\ w_1, & w_2, & w_3 \end{array}$

5.2 Equations of a line in \mathbb{R}^3

A line is always described by two points.

Let us start with a line through the origin and a point $V = (v_1, v_2, v_3)$, i.e., we are considering the line described by the vector $\vec{v} = \overrightarrow{OV} = (v_1, v_2, v_3)$ and so all the points in this line ℓ are described by

$$(x, y, z) = t\vec{v} = (tv_1, tv_2, tv_3) \quad \forall t \in \mathbb{R} \iff \begin{cases} x = tv_1 \\ y = tv_2 \\ z = tv_3 \end{cases}$$

A general line in the space is then described by a vector \vec{v} (giving the direction) and a point $C = (x_C, y_C, z_C)$ and the parametric equations are:

$$\begin{cases} x = x_C + tv_1 \\ y = y_C + tv_2 \\ z = z_C + tv_3 \end{cases} \quad \forall t \in \mathbb{R}$$

The above description of a line is equivalent to describe it giving two points $A = (x_A, y_A, z_A)$ and $B = (x_B, y_B, z_B)$

$$\begin{cases} x = x_A + t(x_B - x_A) \\ y = y_A + t(y_B - y_A) \\ z = z_A + t(z_B - z_A) \end{cases} \iff \begin{cases} x = x_B + t(x_B - x_A) \\ y = y_B + t(y_B - y_A) \\ z = z_B + t(z_B - z_A) \end{cases} \quad \forall t \in \mathbb{R}$$

A line in the space is identified by the intersection between two (non-parallel) planes:

$$\begin{cases} \alpha x + \beta y + \gamma z + \delta = 0 \leftarrow \text{Cartesian equation of a plane} \\ \alpha' x + \beta' y + \gamma' z + \delta' = 0 \leftarrow \text{Cartesian equation of a plane} \end{cases} \quad \alpha, \beta, \gamma, \delta, \alpha', \beta', \gamma', \delta' \in \mathbb{R}$$

6 Vector space

- 0 - dimensional object $\rightarrow \mathbb{R}^0$
- 1 - dimensional object $\rightarrow \mathbb{R}^1$
- 2 - dimensional object $\rightarrow \mathbb{R}^2$
 - 0 - dimensional object (points)
 - 1 - dimensional object (lines)
- 3 - dimensional object $\rightarrow \mathbb{R}^3$
 - 0 - dimensional object (points)
 - 1 - dimensional object (lines)
 - 2 - dimensional object (planes)

Take now 2 cubes move one of them along to a new (the 4th) direction, connect the vertexes and we get a 4-dimensional object and if we extend it at the infinity in all the 4 direction, then we get the 4 - dimensional space $\rightarrow \mathbb{R}^4$. A n -dimensional space is described by \mathbb{R}^n .

In \mathbb{R}^n , a k -dimensional object (where $0 \leq k \leq n$ is described by parametric equations with k different parameter, it has k degrees of freedom. Given a space of dimension n (i.e., \mathbb{R}^n) a k -dimensional object is a subspace of \mathbb{R}^n of dimension k .

$$\mathbb{R}^n = \mathbb{R} \times \mathbb{R} \times \cdots \times \mathbb{R} = \{(x_1, x_2, \dots, x_n) : x_1, x_2, \dots, x_n \in \mathbb{R}\}$$

Similarly to \mathbb{R}^2 and \mathbb{R}^3 , also in \mathbb{R}^n points and vectors are the same objects described by n -tuple of real numbers.

- Multiplication by a scalar

$$\begin{aligned} \mathbb{R} \times \mathbb{R}^n &\longrightarrow \mathbb{R}^n \\ (t, \vec{v}) &\longmapsto t\vec{v} = (tv_1, tv_2, \dots, tv_n) \end{aligned}$$

- Sum (componentwise)

$$\begin{aligned} \mathbb{R}^n \times \mathbb{R}^n &\longrightarrow \mathbb{R}^n \\ (\vec{v}, \vec{w}) &\longmapsto \vec{v} + \vec{w} = (v_1 + w_1, \dots, v_n + w_n) \end{aligned}$$

\mathbb{R}^n equipped with these two operations has nice proprieties! Consider a set V equipped with two operations:

$$\begin{aligned} * : V \times V &\longrightarrow V \\ \odot : \mathbb{R} \times V &\longrightarrow V \end{aligned}$$

where these operations satisfy the following proprieties:

1. $*$ is associative: $\forall u, v, w \in V \quad (u * v) * w = u * (v * w)$
2. $*$ is commutative: $\forall u, v \in V \quad u * v = v * u$
3. existence of identity: $\exists e \in V$ s.t. $\forall v \in V \quad e * v = v$
4. existence of inverses: $\forall v \in V, \exists w \in V$ s.t. $v * w = e$
5. $\forall v \in V, 1 \odot v = v$
6. $\forall s, t \in \mathbb{R}, \forall v \in V, \quad (s + t) \odot v = (s \odot v) * (t \odot v)$
7. $\forall s \in \mathbb{R}, \forall v, w \in V \quad t \odot (v * w) = (t \odot v) * (t \odot w)$

Then we say that V is a vector space over \mathbb{R} (real vector space)

Remark

\mathbb{R}^n with our operation is a vector space over \mathbb{R}

In a vector space V we always have a basis that is a set of elements such that all the elements of V are linear combination of the elements of the basis.

In \mathbb{R}^n , a basis contains always n elements, indeed a set of n linear independent vectors generates all the elements of \mathbb{R}^n .

In other words if $\{\vec{b}_1, \vec{b}_2, \dots, \vec{b}_n\}$ is a basis of \mathbb{R}^n , then $\forall \vec{v} \in \mathbb{R}^n$ we have $\vec{v} = a_1\vec{b}_1 + a_2\vec{b}_2 + \dots + a_n\vec{b}_n$ for $a_1, \dots, a_n \in \mathbb{R}$.

A special basis is the canonical basis $\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n\}$:

$$\vec{e}_1 = (1, 0, \dots, 0) \quad \vec{e}_2 = (0, 1, \dots, 0) \quad \dots \quad \vec{e}_{n-1} = (0, \dots, 1, 0) \quad \vec{e}_n = (0, \dots, 0, 1)$$

In \mathbb{R}^n , we say that m vectors $\vec{v}_1, \dots, \vec{v}_m$ are linearly independent when

$$c_1\vec{v}_1 + \dots + c_m\vec{v}_m = \vec{0} \iff c_1 = \dots = c_m = 0$$

We say that they are linearly dependent when there exist $c_1, \dots, c_m \in \mathbb{R}$ not all zeros such that $c_1\vec{v}_1 + \dots + c_m\vec{v}_m = \vec{0}$

In \mathbb{R}^n , we can construct a basis if we take n linearly independent vectors.

6.1 Euclidean norm

Given a vector $\vec{v} \in \mathbb{R}^n$ the Euclidean norm of \vec{v} is

$$\begin{aligned} \|\vec{v}\| &= d(\vec{v}, \vec{0}) = \sqrt{v_1^2 + v_2^2 + \dots + v_n^2} \\ &= d(V, O) \end{aligned}$$

The Euclidean norm satisfies the triangle inequality:

$$\forall \vec{v}, \vec{w} \in \mathbb{R}^n \quad \|\vec{v} + \vec{w}\| \leq \|\vec{v}\| + \|\vec{w}\|$$

6.2 Dot product

$$\begin{aligned} \bullet : \mathbb{R}^n \times \mathbb{R}^n &\longrightarrow \mathbb{R} \\ (\vec{v}, \vec{w}) &\longmapsto \vec{v} \bullet \vec{w} = v_1w_1 + v_2w_2 + \dots + v_nw_n \end{aligned}$$

In \mathbb{R}^2 and \mathbb{R}^3 (in general in \mathbb{R}^n two vectors \vec{v} and \vec{w} are orthogonal if and only if $\vec{v} \bullet \vec{w} = 0$)

6.3 Equation of a plane given an orthogonal vector and a point

Considering a vector $\vec{n} = (n_1, n_2, n_3) \in \mathbb{R}^3$, the plane through the origin and orthogonal to \vec{n} is the set of points (x, y, z) such that the vector (x, y, z) is orthogonal to \vec{n} , i.e., they must satisfy

$$(x, y, z) \bullet (n_1, n_2, n_3) = 0$$

\Updownarrow

$$n_1x + n_2y + n_3z = 0$$

Cartesian equation of the plane through the origin and orthogonal to \vec{n} .

In general, the equation of a plane through a point $A = (x_A, y_A, z_A) \in \mathbb{R}^3$ and orthogonal to \vec{n} in the set of points $(x, y, z) \in \mathbb{R}^3$ such that the vector $(x - x_A, y - y_A, z - z_A)$ is orthogonal to \vec{n} and so

$$(x - x_A, y - y_A, z - z_A) \bullet (n_1, n_2, n_3) = 0$$

\Updownarrow

$$n_1(x - x_A) + n_2(y - y_A) + n_3(z - z_A) = 0$$

Cartesian equation of a plane through the point A and orthogonal to \vec{v} .

Remark

Given $X \subseteq \mathbb{R}^n$, in order to check if X is a vector space, it is sufficient to check if ($X \neq \emptyset$) it is closed under addition and multiplication by a scalar, i.e,

$$\forall x, y \in X, \text{ check if } x + y \in X$$

$$\forall x \in X, \forall \lambda \in \mathbb{R}, \text{ check if } \lambda x \in X$$

7 Matrices

Given $m, n \in \mathbb{N} \setminus \{0\}$ an $m \times n$ matrix (m rows and n columns). Is and object of this kind

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{bmatrix}$$

where the elements of A are $a_{ij} \in \mathbb{R} \quad \forall i, j \quad 1 \leq i \leq m \quad 1 \leq j \leq n$.

Sometimes we will also write $\bar{A} = (a_{ij})$. We also call the elements a_{ij} as element of the matrix. A matrix is an element of $\mathbb{R}^{m \times n} = \mathbb{R}^{m \cdot n} = \mathbb{R}^{m,n}$

If a matrix $A \in \mathbb{R}^{m \times m}$ is called a square matrix

A matrix in $\mathbb{R}^{1 \times n}$ is called a row vector

A matrix in $\mathbb{R}^{m \times 1}$ is called a column vector

Since $\mathbb{R}^{m \times n}$ (the set of all matrices with m rows and n columns) is a set of the kind \mathbb{R}^K , we know that $\mathbb{R}^{m \times n}$ is a vector space where the operations are

- Addition (componentwise)

$$\forall A, B \in \mathbb{R}^{m \times n} \quad \text{we have } A + B = (a_{ij} + b_{ij}) \in \mathbb{R}^{m \times n}$$

$$\begin{bmatrix} a_{11} + b_{11} & a_{12} + b_{12} & \dots & a_{1m} + b_{1m} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} + b_{m1} & a_{m2} + b_{m2} & \dots & a_{mn} + b_{mn} \end{bmatrix}$$

- Multiplication by a scalar

$\forall A \in \mathbb{R}^{m \times n} \quad \forall \lambda \in \mathbb{R}$ we have $\lambda A = (\lambda a_{ij}) \in \mathbb{R}^{m \times n}$

$$\begin{bmatrix} \lambda a_{11} & \lambda a_{12} & \dots & \lambda a_{1m} \\ \vdots & \vdots & \ddots & \vdots \\ \lambda a_{m1} & \lambda a_{m2} & \dots & \lambda a_{mn} \end{bmatrix}$$

In general a linear system is a set of equation with a certain number of unknowns that must satisfy these equations simultaneously:

$$\left\{ \begin{array}{l} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \vdots \qquad \vdots \qquad \ddots \qquad \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m \end{array} \right.$$

where x_1, x_2, \dots, x_n are unknowns

$a_{ij} \in \mathbb{R}$ given real numbers, coefficients of the linear system where $1 \leq i \leq m$ and $1 \leq j \leq n$

$b_k \in \mathbb{R}$ given real numbers, constant term of the linear system where $1 \leq k \leq m$

Given the above linear system, we can associate to it some matrices:

The matrix of coefficients $A = (a_{ij}) \in \mathbb{R}^{m \times n}$

The vector (matrix) of unknowns $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^{n \times 1}$

The vector (matrix) of constant terms $\vec{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix} \in \mathbb{R}^{m \times 1}$

Linear system with m equations and n unknowns.

The definition of product became, thinking a linear system with

matrix of coefficients $A \in \mathbb{R}^{m \times n}$

vector of unknowns $\vec{x} \in \mathbb{R}^{n \times 1}$

vector of constant terms $\vec{b} \in \mathbb{R}^{m \times 1}$

is equivalent to

$$A\vec{x} = \vec{b}$$

$$\mathbb{R}^{m \times n} \times \mathbb{R}^{n \times 1} \rightarrow \mathbb{R}^{m \times 1}$$

Remark

The dot product between two vectors $\vec{v} = (v_1, \dots, v_n)$ and $\vec{w} = (w_1, \dots, w_n)$ can be written also

$$\begin{bmatrix} v_1, \dots, v_n \end{bmatrix} \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix} = v_1w_1 + v_2w_2 + \dots + v_nw_n = \vec{v} \bullet \vec{w} \rightarrow \mathbb{R}$$

Definition

The transpose of a matrix $A \in \mathbb{R}^{m \times n}$ is a new matrix

$$A^T \in \mathbb{R}^{n \times m}$$

where the rows of A^T are the columns of A or equivalently the columns of A^T are the rows of A

$$A = \begin{bmatrix} 1 & \pi & -\sqrt{2} & 3 \\ 0 & -1 & 0 & 3 \\ \sqrt{3} & \frac{1+\sqrt{5}}{2} & 2\pi & -1 \end{bmatrix} \quad A^T = \begin{bmatrix} 1 & 0 & \sqrt{3} \\ \pi & -1 & \frac{1+\sqrt{5}}{2} \\ -\sqrt{2} & 0 & 2\pi \\ 3 & 3 & -1 \end{bmatrix}$$

Remark

The product between matrices is non commutative!

When A and B are square matrices, e.g., in $\mathbb{R}^{n \times n}$ then

$$AB \in \mathbb{R}^{n \times n} \quad \text{and} \quad BA \in \mathbb{R}^{n \times n} \quad \text{but usually}$$

$$AB \neq BA$$

Example

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \quad B = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$AB = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \quad BA = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

In the example $AB = 0$ product is 0 even if $A, B \neq 0$

$$A + \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = A$$

in general if $A \in \mathbb{R}^{m \times n}$, then $0 \in \mathbb{R}^{m \times n}$ whole element are all zeros is the identity.

Definition

The matrix $I_n \in \mathbb{R}^{n \times n}$ with $I_n = \begin{bmatrix} 1 & \cdots & 0 \\ \vdots & 1 & \vdots \\ 0 & \cdots & 1 \end{bmatrix}$

$$I_n = (I_{ij}) \quad \begin{aligned} I_{ij} &= 0 \text{ if } i \neq j \\ I_{ii} &= 1 \text{ for all } i = 1, \dots, n \end{aligned}$$

is the **identity matrix**. Indeed $\forall A \in \mathbb{R}^{n \times n}$ we have $AI_n = I_nA = A$

8 Solving linear systems

Definition

Two linear system are equivalent if they have the same solutions

$$\begin{cases} x + 2y + z = 1 \\ 3x - 4y + z = 0 \end{cases} \quad \text{is equivalent to} \quad \begin{cases} x + 2y + z = 1 \\ 2x - y + z = \frac{1}{2} \end{cases}$$

First of all, we focus on homogeneous linear system, which are linear system where the constant term are all zeros:

$$A\vec{x} = \vec{0}$$

Our goal is to introduce some «transportation» which are called elementary operations on the matrix A such that the new matrix B where

$$B\vec{x} = \vec{0} \quad \text{is equivalent to} \quad A\vec{x} = \vec{0}$$

Let us define there elementary operations $A \in \mathbb{R}^{m \times n}$

1. **Row switching:** switch two rows of A (if we switch two rows of A this is equivalent to switch two equations of the linear system $A\vec{x} = \vec{0}$; we get an equivalent new linear system).

We can switch two rows of A performing the following product:

$$S_{ij}A$$

where S_{ij} is a matrix in $\mathbb{R}^{m \times m}$ obtained from the identity matrix I_m switching the i^{th} row of I_m with the j^{th} row of I_m

$$A\vec{x} = \vec{0}$$

\Updownarrow

$$S_{ij}A\vec{x} = S_{ij}\vec{0}$$

$$\Updownarrow \\ B\vec{b} = \vec{0}$$

where B is obtained from A switching the i^{th} row with the j^{th} row

2. **Row multiplication:** Multiply a row of A by a scalar $\lambda \in \mathbb{R}$ (if you multiply an equation by a scalar you get an equivalent equation)

We can obtain this operation, considering the following product

$$D_i(\lambda)A$$

Where $D_i(\lambda)$ is a diagonal matrix $m \times m$ obtained from I_m replacing I_{ii} with λ

$$A\vec{x} = \vec{0} \iff D_i(\lambda)A\vec{x} = \vec{0} \iff B\vec{x} = \vec{0}$$

3. **Row addition:** add to a row of A another row of A multiplied by a scalar $\lambda \in \mathbb{R}$. We can obtain this elementary operation, considering the following product.

$$E_{ij}(\lambda)A$$

Where $E_{ij}(\lambda)$ is a $m \times m$ matrix obtained from I_m replacing the element I_{ij} with $\lambda \in \mathbb{R}$. This product replaces the i^{th} row of A with the sum of i^{th} row of A with the j^{th} row of A multiplied by λ .

$$A\vec{x} = \vec{0} \iff E_{ij}(\lambda)A\vec{x} = \vec{0} \iff B\vec{x} = \vec{0}$$

Definition

A square matrix $A \in \mathbb{R}^{n \times n}$ is invertible (w.r.t. the products) if there exist a matrix $B \in \mathbb{R}^{n \times n}$ such that $AB = BA = I_n$

In this case we write $B = A^{-1}$. The inverse of A is denoted by A^{-1}

Since when we start from a linear system with constant vector \vec{b} we get an equivalent linear system with a new constant vector, it is useful to def with the augmented matrix $(A|b) \in \mathbb{R}^{m \times n+1}$

In general, in order to solve a linear system $A\vec{x} = \vec{b}$ we want to obtain an equivalent linear system $B\vec{x} = \vec{c}$ where

$$(B|\vec{c}) \quad (\text{obtained from } (A|\vec{b}) \text{ using elementary operations})$$

is in the so-called **Row-echelon form**.

Definition

A matrix $M \in \mathbb{R}^{a \times b}$ is in **row-echelon form (REF)** if:

1. The leading (first nonzero) entry of each nonzero row (called a *pivot*) is strictly to the right of the pivot of the row above it.
2. All zero rows (if any) are at the bottom of the matrix.

The rows of a matrix, denoted by $\text{rank}(M)$, is the number of pivot of any REF for M . A column of M is called pivot column if it contains one pivot.

A matrix M is in reduced row-echelon form RREF if

1. in in REF
2. all the pivot are 1
3. the only non-zero element of any pivot column is the pivot itself (REF is not unique. RREF is unique)

Let consider $A\vec{x} = \vec{b}$ where $A \in \mathbb{R}^{m \times n}$ and focus on the augmented matrix $(A|\vec{b})$, then we transform it into a matrix $(A'|\vec{b}')$ that is REF and $A'\vec{x} = \vec{b}'$ is equivalent to $A\vec{x} = \vec{b}$

1. If $\text{rank}(A'|\vec{b}') \neq \text{rank}(A')$, then $A'\vec{x} = \vec{b}'$ and $A\vec{x} = \vec{b}$ have no solutions

$$A' = \begin{bmatrix} 1 & -1 & 2 \\ 0 & 5 & 3 \\ 0 & 0 & -2 \\ 0 & 0 & 0 \end{bmatrix} \quad \text{rank}(A') = 3$$

$$(A'|\vec{b}') = \left[\begin{array}{ccc|c} 1 & -1 & 2 & 0 \\ 0 & 5 & 3 & -1 \\ 0 & 0 & -2 & 2 \\ 0 & 0 & 0 & 5 \end{array} \right] \quad \text{rank}(A'|\vec{b}') = 4 \neq \text{rank}(A')$$

$$\left\{ \begin{array}{l} x - y + 2z = 0 \\ 5y + 3z = -1 \\ -2z = 2 \\ 0 = 5 \end{array} \right.$$

Indeed, this means that the column \vec{b}' in the augmented matrix must contain a pivot, i.e., a non-zero element k in correspondence to a zero row of A' and this yields to a false identity of the kind

$$0 = k \quad k \in \mathbb{R} \quad k \neq 0$$

2. If $\text{rank}(A'|\vec{b}') = \text{rank}(A') = n$ number of unknowns, then there exists a unique

solution

$$A' = \begin{bmatrix} 1 & -1 & 2 \\ 0 & 5 & 3 \\ 0 & 0 & -2 \\ 0 & 0 & 0 \end{bmatrix} \quad A' \in \mathbb{R}^{4 \times 3} \Rightarrow 4 \text{ equations, 3 unknowns}$$

$$\text{rank}(A') = 3$$

$$(A'|\vec{b}') = \left[\begin{array}{ccc|c} 1 & -1 & 2 & -1 \\ 0 & 5 & 3 & 0 \\ 0 & 0 & -2 & 3 \\ 0 & 0 & 0 & 0 \end{array} \right] \quad \text{rank}(A'|\vec{b}') = 3$$

$$\left\{ \begin{array}{l} x - y + 2z = -1 \\ 5y + 3z = 0 \\ -2z = 3 \\ 0 = 0 \end{array} \right.$$

3. If $\text{rank}(A'|\vec{b}') = \text{rank}(A') < n$ (n = number of unknowns) the linear system has infinitely many solutions.

Consider for example

$$A' = \begin{bmatrix} 1 & -1 & 2 \\ 0 & 5 & 3 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \text{rank}(A') = 2 < 3 = n$$

$$(A'|\vec{b}') = \left[\begin{array}{ccc|c} 1 & -1 & 2 & \sqrt{2} \\ 0 & 5 & 3 & -1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \quad \text{rank}(A'|\vec{b}') = 2$$

\Updownarrow

$$\left\{ \begin{array}{l} x - y + 2z = \sqrt{2} \\ 5y + 3z = -1 \\ 0 = 0 \\ 0 = 0 \end{array} \quad \forall z \in \mathbb{R} \right.$$

we have infinitely many solutions, because z is a free variable, it can assume any value in \mathbb{R}

Rouché–Capelli Theorem

Let $A\vec{x} = \vec{b}$ be a linear system, with $A \in \mathbb{R}^{m \times n}$ ($\begin{smallmatrix} n \text{ equations} \\ m \text{ unknowns} \end{smallmatrix}$) and let $(A'|\vec{b}')$ be in REF obtained from $(A|\vec{b})$, then (this $A'\vec{x}' = \vec{b}'$ is equivalent to $A\vec{x} = \vec{b}$)

1. if $\text{rank}(A|\vec{b}) \neq \text{rank}(A')$, then $A'\vec{x} = \vec{b}$ doesn't have solutions
2. if $\text{rank}(A|\vec{b}) = \text{rank}(A') = n$ then $A'\vec{x} = \vec{b}$ has a unique solution
3. if $\text{rank}(A|\vec{b}) = \text{rank}(A') < n$ then $A'\vec{x} = \vec{b}$ has infinitely many solutions and $n - \text{rank}(A')$ is the number of free variables

9 Vector subspaces

Given a vector space V and $W \subseteq V$, we say that W is a vector subspace of V if it is a vector space.

Given a vector space V and $W \subseteq V$ then W is a vector subspace if and only if

- a) said $e \in V$ the identity w.r.t. the «addition», we have $e \in W$
- b) $\forall w_1, w_2 \in W$, we have $w_1 * w_2 \in W$
- c) $\forall w \in W, \forall \lambda \in \mathbb{R}$, we have $\lambda \odot w \in W$

Given a homogeneous linear system, the set of solutions (if it is not the empty set) is a vector space (in particular is a vector subspace of \mathbb{R}^n where n is the number of unknowns)

Consider a linear system $A\vec{x} = \vec{0}$ where $A \in \mathbb{R}^{m \times n}$, let S be the set of solutions. We can have that $S = \{(0, \dots, 0) \in \mathbb{R}^n\} \subseteq \mathbb{R}^n$ and it is a vector space (the trivial one). In general

we have that $S = \left\{ \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n = \mathbb{R}^{n \times 1} : A\vec{x} = \vec{0} \right\}$

- a) We can check that $\vec{0} \in S$
- b) $\forall \vec{v}, \vec{w} \in S$, i.e., $A\vec{v} = \vec{0}$ and $A\vec{w} = \vec{0}$ we want to check that $\vec{v} + \vec{w} \in S$, i.e., we want to check that

$$A(\vec{v} + \vec{w}) = \vec{0}$$

indeed we can observe that $A(\vec{v} + \vec{w}) = A\vec{v} + A\vec{w} = \vec{0} + \vec{0} = \vec{0}$

- c) $\forall \vec{v} \in S, \forall \lambda \in \mathbb{R}$ we check that $\lambda\vec{v} \in S$, i.e., $A(\lambda\vec{v}) = \vec{0}$

Given a non-homogeneous linear system, the set of its solutions is not a vector space.

Let us consider $A\vec{x} = \vec{b}$ where $A \in \mathbb{R}^{m \times n}$ and $\vec{b} \neq \vec{0}$ then $\vec{0} \notin S = \left\{ \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n : A\vec{x} = \vec{b} \right\}$
because $A\vec{0} = \vec{0} \neq \vec{b}$

Definition

Let V be a vector space and $v_1, v_2, \dots, v_k \in V$ we define the set generated by v_1, \dots, v_k , and we call it the spanning set of v_1, \dots, v_n the following set

$$\text{span}(v_1, \dots, v_k) = \langle v_1, \dots, v_k \rangle = \{c_1v_1 + \dots + c_kv_k : \forall c_1, \dots, c_k \in \mathbb{R}\}$$

Example

In \mathbb{R}^3 , given two vectors \vec{v}_1, \vec{v}_2 linearly independent, their spamming set is a plane through the origin

$$\text{span}(\vec{v}_1, \vec{v}_2) = \{s\vec{v}_1 + t\vec{v}_2 : \forall s, t \in \mathbb{R}\}$$

Definition

Let V be a vector space, we say that $\{v_1, \dots, v_k\}$ with $v_1, \dots, v_k \in V$ is set of generators of V if $V = \text{span}(v_1, \dots, v_k)$

Definition

Let V be a vector space, we say that $\{b_1, \dots, b_n\}$ with $b_1, \dots, b_n \in V$ is a basis of V if

- $V = \text{span}(b_1, \dots, b_n)$
- b_1, \dots, b_n are linearly independent

Definition

Let V be a vector space, we sat that $w_1, \dots, w_h \in V$ are linearly independent if (given $c_1, \dots, c_h \in \mathbb{R}$)

$$c_1w_1 + \dots + c_hw_h = 0 \quad (1)$$

\Updownarrow

$$c_1 = \dots = c_h = 0$$

if (1) has a solution c_1, \dots, c_h not all zeros, then w_1, \dots, w_h are linearly dependent.

Given a vector $\vec{v} \in \mathbb{R}^4$, the component w.r.t the canonical bases are (v_1, v_2, v_3, v_4) because

$$\vec{v} = v_1\vec{e}_1 + v_2\vec{e}_2 + v_3\vec{e}_3 + v_4\vec{e}_4$$

that is v_1, v_2, v_3, v_4 are the coefficent og the linear combination of $\vec{e}_1, \vec{e}_2, \vec{e}_3, \vec{e}_4$ from which we obtain \vec{v} .

If we have another basis $\vec{b}_1, \vec{b}_2, \vec{b}_3, \vec{b}_4$, then the component og \vec{v} are (w_1, w_2, w_3, w_4) such that

$$\vec{v} = w_1 \vec{b}_1 + w_2 \vec{b}_2 + w_3 \vec{b}_3 + w_4 \vec{b}_4$$

In the previous example the component of $(1, 0, 3, 5)$ w.r.t. the canonical basis are $(1, 0, 3, 5)$, but the component w.r.t. the basis $(1, 2, 3, 5), (2, 1, 1, 0), (1, 1, 1, 0), (0, 1, 0, 0)$ are

$$(1, 0, 3, 5) = 1 \cdot (1, 2, 3, 5) + 0 \cdot (2, 1, 1, 0) + 0 \cdot (1, 1, 1, 0) + (-2)(0, 1, 0, 0)$$

Two linear system are equivalent if the sets of solutions are the same

Let S_1 be the set of solutions of $\begin{cases} x_1 - x_2 + 7x_3 = 0 \\ x_1 = x_2 - 7x_3 \end{cases}$

$$S_1 = \{(x_2 - 7x_3, x_2, x_3, x_4) : x_2, x_3, x_4 \in \mathbb{R}\}$$

$(-6, 1, 1, 5)$ is a solution of the linear system

Let S_2 be the set of solutions of $\begin{cases} -2x_1 + 2x_2 - 14x_3 + 5x_4 = 0 \\ x_4 = 0 \end{cases}$

$$S_2 = \{(x_2 - 7x_3, x_2, x_3, 0) : x_2, x_3 \in \mathbb{R}\}$$

$$S_2 \subseteq S_1 \quad \text{but} \quad S_2 \neq S_1$$

$$S_1, S_2 \subseteq \mathbb{R}^4 \quad \begin{aligned} S_1 &\text{ has 3 free variables} \\ S_2 &\text{ has 2 free variables} \end{aligned}$$

$$c_1 \vec{b}_1 + c_2 \vec{b}_2 + c_3 \vec{b}_3 + c_4 \vec{b}_4 = \vec{v}$$

Let V be a vector setspace

Definition

We say that $\{v_1, \dots, v_k\}$, whith $v_1, \dots, v_k \in V$ is a set of generators of V if

$$V = \text{span}(v_1, \dots, v_k) = \{a_1 v_1 + \dots + a_k v_k : a_1, \dots, a_k \in \mathbb{R}\}$$

That is $\forall v \in V$, we have $v = c_1 v_1 + \dots + c_k v_k$ for some $c_1, \dots, c_k \in \mathbb{R}$

Definition

We say that $\{b_1, \dots, b_h\}$, with $b_1, \dots, b_h \in V$ is a basis of V if

1. $V = \text{span}(b_1, \dots, b_h)$ (i.e., b_1, b_h is a set of generators)
2. b_1, \dots, b_h are linearly idependet

Example

$\{\vec{e}_1, \vec{e}_2, \vec{e}_3\}$ is a basis of \mathbb{R}^3 , $\{\vec{e}_1, \vec{e}_2, \vec{e}_3, (1, 2, 3)\}$ is a set of generators of \mathbb{R}^3 but not a basis

Definition

The number of elements in any basis of V is called the dimension of V

The dimension of \mathbb{R}^n is n

Given $W \subseteq \mathbb{R}^n$ s.t. W is a vector space, i.e., a vector subspace of \mathbb{R}^n , then the dimension W can be $1, 2, \dots, n - 1$ (where W is a proper subset of \mathbb{R}^n). If $W = \{\vec{0}\}$ then it is a trivial vector subspace of \mathbb{R}^n with dimension 0.

In \mathbb{R}^2 we can have

- dimension $W = 0$, $W = \{(0, 0)\}$
- dimension $W = 1$, then a basis is $\{\vec{v}\}$ and $W = \text{span}(\vec{v}) = \{t\vec{v} : t \in \mathbb{R}\}$ that is W is a line in \mathbb{R}^2 containing the origin

In \mathbb{R}^3 , we can have

- dimension $W = 0$, $W = \{(0, 0, 0)\}$
- dimension $W = 1$, then a basis is $\{\vec{v}\}$ and $W = \text{span}(\vec{v}, \vec{w})$ that is W is a line in \mathbb{R}^3 containing the origin
- dimension $W = 2$, then a basis is $\{\vec{v}, \vec{w}\}$ and $W = \text{span}(\vec{v}, \vec{w})$ that is W is a plane in \mathbb{R}^3 containing the origin.
- dimension $W = h$, for $1 \leq h \leq n - 1$, W is a h -dim. space through the origin described by a basis $\{\vec{b}_1, \dots, \vec{b}_h\}$, i.e. $W = \text{span}(\vec{b}_1, \dots, \vec{b}_h)$

Given two vector spaces V and W , we say that they are isomorphic if there exist a function (called isomorphism)

$$f : V \rightarrow W$$

s.t.

1. f is surjective: every element of W has pre-image, i.e., $\forall w \in W$ there exists $v \in V$ s.t. $f(v) = w$
2. f is injective: $\forall v_1, v_2 \in V$ if $v_1 \neq v_2$ then $f(v_1) \neq f(v_2)$ equivalently we can say that $\forall v_1, v_2 \in V$

$$f(v_1) = f(v_2) \Leftrightarrow v_1 = v_2$$

(A function f is both injective and surjective is called a bijection and it is a one-to-one correspondence)

3. Let $+_v$ and $+_w$ be the operation between elements of V and W , respectively, $\forall v_1, v_2 \in V$
- $$f(v_1 +_v v_2) = f(v_1) +_w f(v_2)$$
- $$\forall k \in \mathbb{R}, \forall v \in V, f(kv) = kf(v)$$

A function f only satisfying 3. is called a linear map (a bijective linear map is an isomorphism).

A vector space V of dimension n is isomorphic to \mathbb{R}^n indeed, let $\{b_1, \dots, b_n\}$ be a basis of V , this $\forall v \in V$

$$v = x_1 b_1 + x_2 b_2 + \dots + x_n b_n \quad \text{for } x_1, x_2, \dots, x_n \in \mathbb{R}$$

then

$$\begin{aligned} f : V &\rightarrow \mathbb{R}^n \\ v &\mapsto (x_1, x_2, \dots, x_n) \end{aligned}$$

is an isomorphism.

9.1 Example of vector space with infinite dimension

Let $\mathbb{R}[x]$ be the set of polynomials with coefficients in \mathbb{R}

$$\mathbb{R}[x] = \{a_0 + a_1 x + a_2 x^2 + a_3 x^3 + \dots + a_n x^n : a_0, \dots, a_n \in \mathbb{R} \quad \forall n \in \mathbb{N}\}$$

Given $p(x) = a_0 + a_1 x + \dots + a_n x^n$ where $a_n \neq 0$ and we say that the degree $\deg p(x)$ is n

Example

$$\begin{aligned} p_1(x) &= 1 + 2x + 3x^3 + x^4 & p_2(x) &= \frac{1}{2} - x + 2x^2 \\ p_1(x) + p_2(x) &= \frac{3}{2} + x + 2x^2 + 3x^3 + x^4 \end{aligned}$$

Example

$$\begin{aligned} p(x) &= -3 + x + \sqrt{2}x^2 + x^5 - 7x^6, & k &= 2 \\ kp(x) &= -6 + 2x + 2\sqrt{2}x^2 + 2x^5 - 14x^6 \end{aligned}$$

The set $\mathbb{R}[x]$ with these operations is a vector space.

There is a one to one correspondence (an injective and surjective map) between $\mathbb{R}[x]$ and \mathbb{R}^∞ what is the set of vectors of infinite length

$$\begin{aligned} f : \mathbb{R}[x] &\rightarrow \mathbb{R}^\infty \\ p(x) &\mapsto (a_0, a_1, \dots, a_n, 0, 0, 0, \dots) \end{aligned}$$

This map is also a linear map (meaning that $f(p_1(x)+p_2(x)) = f(p_1(x))+f(p_2(x))$, $f(kp(x)) = kf(p(x))$)

This f is an isomorphism and $\mathbb{R}[x]$ is «equal» (up to isomorphism) to \mathbb{R}^∞ .

A basis of \mathbb{R}^∞ is

$$\{(1, 0, 0, \dots), (0, 1, 0, \dots), (0, 0, 1, 0, 0, \dots), \dots\}$$

and the dimension of \mathbb{R}^∞ is ∞

$$p(x) = a_0 \cdot 1 + a_1 \cdot x + a_2 \cdot x^2 + \dots + a_n \cdot x^n + 0 \cdot x^{n+1} + \dots$$

10 Determinant of a matrix

Consider $A\vec{x} = \vec{b}$ for solving it we transform using elementary operations the augmented matrix $(A|\vec{b})$ in RREF.

If $A \in \mathbb{R}^{n \times n}$, we can solve $A\vec{x} = \vec{b}$ also in another way

$$\begin{aligned} 5x &= 3 \\ x &= \frac{1}{5} \cdot 3 \end{aligned}$$

if $A \in \mathbb{R}^{n \times n}$ is an invertible matrix, that is there exists a matrix $B \in \mathbb{R}^{n \times n}$ such that $AB = BA = I_n$ and we denote B by A^{-1} and we call it the inverse of A . This if A is invertible and the inverse in A^{-1} then

$$A\vec{x} = \vec{b} \Leftrightarrow A^{-1} \cdot A\vec{x} = A^{-1}\vec{b} \Leftrightarrow I_n \cdot \vec{x} = A^{-1}\vec{b} \Leftrightarrow \vec{x} = A^{-1}\vec{b}$$

For studying the invertibility of a square matrix we introduce the notion of determinant.

The determinant is a real number associated to a square matrix. Let us start with a matrix $A \in \mathbb{R}^{2 \times 2}$ which is

$$\begin{aligned} A &= \begin{bmatrix} a & b \\ c & d \end{bmatrix} \quad a, b, c, d \in \mathbb{R} \\ \det(A) &= ad - bc \end{aligned}$$

Consider now $A \in \mathbb{R}^{3 \times 3}$ which is

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$\det(A) = a_{11} \cdot \det \begin{bmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{bmatrix} - a_{12} \cdot \det \begin{bmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{bmatrix} + a_{13} \cdot \det \begin{bmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix}$$

In general given $A \in \mathbb{R}^{n \times n}$, $A = (a_{ij})$

$$\det A = a_{11} \cdot \det(A_{11}) - a_{12} \cdot \det(A_{12}) + a_{13} \cdot \det(A_{13}) - a_{14} \cdot \det(A_{14}) + \dots + (-1)^n a_{1n} \det(A_{1n})$$

where $A_{11} \in \mathbb{R}^{n-1 \times n-1}$ obtained from A removing first row and first column

where $A_{12} \in \mathbb{R}^{n-1 \times n-1}$ obtained from A removing first row and second column

where $A_{1j} \in \mathbb{R}^{n-1 \times n-1}$ obtained from A removing first row and j^{th} column for $j = 1, \dots, n$. This is the so-called Laplace expansion of the determinat.

Remark

If $A \in \mathbb{R}^{2 \times 2}$, then $\det(A)$ is the area of the parallelogram with sides the columns of A . If $A \in \mathbb{R}^{3 \times 3}$, then $\det(A)$ is the volume of the parallelepiped with sides the columns of A . Similarly this works if $A \in \mathbb{R}^{n \times n}$, $\det(A)$ is the volume of the parallelepiped with dimension n and sides the columns of A .

Remark

A matrix $A \in \mathbb{R}^{n \times n}$ is invertible if and only if $\det(A) \neq 0$

Remark

Consider $A\vec{x} = \vec{b}$ $A \in \mathbb{R}^{n \times n}$, the solution is unique if and only if $\det(A) \neq 0$ and the solution is given by $x_i = \frac{\det(A_i)}{\det(A)}$ where $A_i \in \mathbb{R}^{n \times n}$ obtained from A replacing the i^{th} column of A with \vec{b}

This is the Cramer rule.

Example

$$A = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 2 & 4 \\ 0 & 0 & 7 \end{bmatrix} \quad B = \begin{bmatrix} 0 & 1 & 2 \\ 1 & 1 & 4 \\ 1 & 0 & 0 \end{bmatrix}$$

$$\begin{aligned} \det(A) &= 1 \cdot \det \begin{bmatrix} 2 & 4 \\ 0 & 7 \end{bmatrix} - 0 \det \begin{bmatrix} 0 & 4 \\ 0 & 7 \end{bmatrix} + 2 \det \begin{bmatrix} 0 & 2 \\ 0 & 0 \end{bmatrix} \\ &= 1 \times 14 - 0 + 2 \times 0 = 14 \end{aligned}$$

$$\det(B) = -1 \cdot \det \begin{bmatrix} 1 & 4 \\ 1 & 0 \end{bmatrix} + 2 \det \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} = 4 - 2 = 2$$

Remark

For computing the determinat we can apply the definition exploiting any row (not necessarily the firt one).

This, if a matrix has a row with many zeros is more convenient to use.

Example

$$B = \begin{bmatrix} 0 & 1 & 2 \\ 1 & 1 & 4 \\ 1 & 0 & 0 \end{bmatrix}$$

$$\det(B) = 1 \cdot \det \begin{bmatrix} 1 & 2 \\ 1 & 4 \end{bmatrix} + 0 + 0 = 2$$

$$C = \begin{bmatrix} 1 & -1 & 2 & 4 \\ 0 & 0 & 0 & 0 \\ -7 & 5 & 1 & 2 \\ 3 & 1 & -5 & 1 \end{bmatrix} \quad \det(C) = 0$$

10.1 Properties

1. If $A \in \mathbb{R}^{n \times n}$ is in REF and the number of pivots is n , then $\det(A)$ is equal to the product of pivots
2. If $A \in \mathbb{R}^{n \times n}$ is in REF and the number of pivots is $< n$, then $\det(A) = 0$
3. Given $A \in \mathbb{R}^{n \times n}$ and let B be a matrix obtained from A switching two rows, then $\det(B) = -\det(A)$ ($\det(A) = -\det(B)$)

$$A = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 2 & 4 \\ 0 & 0 & 7 \end{bmatrix} \quad B = \begin{bmatrix} 0 & 2 & 4 \\ 1 & 0 & 2 \\ 0 & 0 & 7 \end{bmatrix}$$

check that $\det(A) = -\det(B)$

4. Given $A \in \mathbb{R}^{n \times n}$ and let B be a matrix obtained from A multiplying a row of A by a scalar k , then $\det(B) = k \det(A)$

$$A = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 2 & 4 \\ 0 & 0 & 7 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 2 & 4 \\ 0 & 0 & 1 \end{bmatrix}$$

check that $\det(B) = \frac{1}{7} \cdot \det(A)$

5. Given $A \in \mathbb{R}^{n \times n}$ and let B be a matrix obtained from A adding to a row of A a multiple of another row, then $\det(B) = \det(A)$

$$A = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 2 & 4 \\ 0 & 0 & 7 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 1 & 4 \\ 0 & 2 & 4 \\ 0 & 0 & 7 \end{bmatrix}$$

check that $\det(B) = \det(A)$

Remark

Pay attention! If you have a matrix A and transform it replacing, e.g., the first row R_1 with the row $2R_1 + 3R_2$ then we obtain a new matrix B and $\det B = 2 \det A$

10.2 Properties

Given $A, B \in \mathbb{R}^{n \times n}$, $\det(AB) = \det(A)\det(B)$ let us see a method for computing the inverse of a matrix $A \in \mathbb{R}^{n \times n}$ such that $\det A \neq 0$.

We want to compute an unknown matrix $A^{-1} \in \mathbb{R}^{n \times n}$ such that

$$A \cdot A^{-1} = I_n$$

that is

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_{11} & \cdots & x_{1n} \\ \vdots & \ddots & \vdots \\ x_{n1} & \cdots & x_{nn} \end{bmatrix} = \begin{bmatrix} 1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & 1 \end{bmatrix}$$

a_{ij} are known real numbers

x_{ij} are unknown real numbers

$$\begin{cases} a_{11}x_{11} + \cdots + a_{1n}x_{n1} = 1 \\ a_{21}x_{11} + \cdots + a_{2n}x_{n1} = 0 \\ \dots \\ a_{n1}x_{11} + \cdots + a_{nn}x_{n1} = 0 \end{cases} \quad \text{Linear system with } n \text{ unknown and } n \text{ equations}$$

\Updownarrow

Linear system whose augmented matrix is $(A|\vec{e}_1)$ where $\vec{e}_1 = (1, 0, \dots, 0)$

Similarly, if we focus on the second column of A^{-1} , we get a linear system whose unknowns are $(x_{12}, x_{22}, x_{32}, \dots, x_{n2})$ and the augmented matrix is

$$(A|\vec{e}_2) \quad \text{where} \quad \vec{e}_2 = (0, 1, 0, \dots, 0)$$

This, in order to find A^{-1} we need to solve n linear systems whose augmented matrices are

$$(A|\vec{e}_1), (A|\vec{e}_2), \dots, (A|\vec{e}_n)$$

We can solve all these linear systems considering

$$(A|I_n) \leftarrow \det A \neq 0$$

and transforming it into RREF we obtain

$$(I_n|A^{-1})$$

Example

$$\begin{aligned}
 A &= \begin{bmatrix} 1 & 2 \\ 3 & 0 \end{bmatrix} \quad \det A = -6 \neq 0 \Rightarrow A \text{ invertible} \\
 \left[\begin{array}{cc|cc} 1 & 2 & 1 & 0 \\ 3 & 0 & 0 & 1 \end{array} \right] &\rightarrow \left[\begin{array}{cc|cc} 1 & 2 & 1 & 0 \\ 0 & 6 & 3 & -1 \end{array} \right] \quad R_1 \\
 \rightarrow \left[\begin{array}{cc|cc} 3 & 0 & 0 & 1 \\ 0 & 6 & 3 & -1 \end{array} \right] \quad R_1 - 3R_2 &\rightarrow \left[\begin{array}{cc|cc} 1 & 0 & 0 & \frac{1}{3} \\ 0 & 1 & \frac{1}{2} & -\frac{1}{6} \end{array} \right] \quad \frac{1}{3}R_1 \\
 A^{-1} &= \begin{bmatrix} 0 & \frac{1}{3} \\ \frac{1}{2} & -\frac{1}{6} \end{bmatrix} \\
 A \cdot A^{-1} &= \begin{bmatrix} 1 & 2 \\ 3 & 0 \end{bmatrix} \begin{bmatrix} 0 & \frac{1}{3} \\ \frac{1}{2} & -\frac{1}{6} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}
 \end{aligned}$$

Check that $A^{-1} \cdot A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

11 Linear function

Definition

Given two vector spaces V and W , with finite dimension. $V \simeq \mathbb{R}^n$ and $W \simeq \mathbb{R}^m$ a linear function $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a map such that

1. $\forall \vec{v}_1, \vec{v}_2 \in \mathbb{R}^n, f(\vec{v}_1 + \vec{v}_2) = f(\vec{v}_1) + f(\vec{v}_2)$
2. $\forall \vec{v} \in \mathbb{R}^n, \forall k \in \mathbb{R}, f(k\vec{v}) = kf(\vec{v})$

$$\begin{gathered}
 f(\vec{0}) = \vec{0} \\
 \vec{0} \in \mathbb{R}^n \quad \vec{0} \in \mathbb{R}^m
 \end{gathered}$$

By 1. we know that $\forall \vec{v} \in \mathbb{R}^n$, then $f(\vec{0} + \vec{v}) = f(\vec{0}) + f(\vec{v}) = f(\vec{v}) \Rightarrow f(\vec{v}) = f(\vec{0}) + f(\vec{v}) \Rightarrow f(\vec{0}) = \vec{0}$

If we know the image of the vectors of a basis of \mathbb{R}^n , then we are able to compute f an all the vectors of \mathbb{R}^n .

If $B = \{\vec{b}_1, \dots, \vec{b}_n\} \subseteq \mathbb{R}^n$ is a basis of \mathbb{R}^n then, $\forall \vec{v} \in \mathbb{R}^n$, we have that $\vec{v} = x_1\vec{b}_1 + x_2\vec{b}_2 + \dots + x_n\vec{b}_n$ fore some $x_1, \dots, x_n \in \mathbb{R}$ and

$$\begin{aligned}
 f(\vec{v}) &= f(x_1\vec{b}_1 + x_2\vec{b}_2 + \dots + x_n\vec{b}_n) \\
 &= x_1f(\vec{b}_1) + x_2f(\vec{b}_2) + \dots + x_nf(\vec{b}_n)
 \end{aligned}$$

Let $C = \{\vec{c}_1, \dots, \vec{c}_m\} \subseteq \mathbb{R}^m$ be a basis of \mathbb{R}^m , if we know that

$$\begin{aligned} f(\vec{b}_1) &= \vec{w}_1 = a_{11}\vec{c}_1 + a_{21}\vec{c}_2 + \cdots + a_{m1}\vec{c}_m \\ f(\vec{b}_2) &= \vec{w}_2 = a_{12}\vec{c}_1 + a_{22}\vec{c}_2 + \cdots + a_{m2}\vec{c}_m \\ &\quad \dots \\ f(\vec{b}_n) &= \vec{w}_n = a_{1n}\vec{c}_1 + a_{2n}\vec{c}_2 + \cdots + a_{mn}\vec{c}_m \end{aligned}$$

We define

$$A = \begin{bmatrix} f(\vec{b}_1) & f(\vec{b}_2) & \cdots & f(\vec{b}_n) \end{bmatrix}$$

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \in \mathbb{R}^{m \times n}$$

This is the matrix whose columns are the component of $f(\vec{b}_i)$ w.r.t. the basis C of \mathbb{R}^m . This matrix A is the matrix associated to f and «it works like f »

Take $\vec{v} \in \mathbb{R}^n$, $\vec{v} = v_1\vec{b}_1 + \cdots + v_n\vec{b}_n$

$$\begin{aligned} f(\vec{v}) &= v_1 f(\vec{b}_1) + \cdots + v_n f(\vec{b}_n) = v_1 \vec{w}_1 + \cdots + v_n \vec{w}_n = \\ &= v_1 (a_{11}\vec{c}_1 + \cdots + a_{1n}\vec{c}_m) + \cdots + v_n (a_{1n}\vec{c}_1 + \cdots + a_{mn}\vec{c}_m) \\ &= (v_1 a_{11} + \cdots + v_n a_{1n}) \vec{c}_1 + \cdots + (v_1 a_{m1} + \cdots + v_n a_{mn}) \vec{c}_m \\ &= W \in \mathbb{R}^m \end{aligned}$$

where $\vec{w} = A\vec{v} \Leftrightarrow \vec{w} = f(\vec{v})$

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} = \begin{bmatrix} a_{11}v_1 + \cdots + a_{1n}v_n \\ \vdots \\ a_{m1}v_1 + \cdots + a_{mn}v_n \end{bmatrix}$$

$$\mathbb{R}^{m \times n} \quad \mathbb{R}^{n \times 1} \quad \mathbb{R}^{m \times 1}$$

Remark

In general, given W_1 and W_2 vector spaces, then $W_1 \cup W_2$ is not necessarily a vector space. Indeed if we take an element $w_1 \in W_1$ and an element $w_2 \in W_2$, then $w_1, w_2 \in W_1 \cup W_2$ ($w_1, w_2 \notin W_1 \cap W_2$) but $w_1 + w_2$ is not necessarily in $W_1 \cup W_2$.

But, $W_1 \cap W_2$ is a vector space, because consider $t_1, t_2 \in W_1 \cap W_2$ then $t_1 + t_2 \in W_1$ ($t_1, t_2 \in W_1$ and W_1 vector space), $t_1 + t_2 \in W_2$ ($t_1, t_2 \in W_2$ and W_2 vector space).

This $t_1 + t_2 \in W_1 \cap W_2$, similar obserbation for multuplation by a scalar.

In our special case, we need to prove (or disprove) that $W_1 \cup W_2$ is closed under addition and multiplication by a scalar. Since W_1 and W_2 are vector spaces we only need to check what happen to $w_1 + w_2$ when

$$w_1 \in W_1 \setminus W_2$$

$$\begin{aligned}
w_2 &\in W_2 \setminus W_1 \\
w_1 &= (y+z, y, z) \quad w_2 = (t, -t, 2t) \\
w_1 + w_2 &= (y+z+t, y-t, z+2t) \in W_1
\end{aligned}$$

Actually, we can prove that $W_2 \subseteq W_1$.

- $(t, -t, 2t)$ a general element of W_2 is also an element of W_1 indeed $-t + 2t = t$
- $W_2 = \text{span}(1, -1, 2)$ and $(1, -1, 2) = -(1, 1, 0) + 2(1, 0, 1)$
 $\Rightarrow (1, -1, 2) \in \text{span}((1, 1, 0), (1, 0, 1))$

$$\begin{aligned}
t(1, -1, 2) &= t(-(1, 1, 0) + 2(1, 0, 1)) = \\
&= -t(1, 1, 0) + 2t(1, 0, 1) \in \text{span}((1, 1, 0), (1, 0, 1)) \in \mathbb{R}^3
\end{aligned}$$

$W_3 = \{(t, 2t, 2t) \in \mathbb{R}^3 : \forall t \in \mathbb{R}\} \subseteq \mathbb{R}^3$ is a vector space but $W_1 \cup W_3$ is not a vector space.
Indeed,

$$(y+z, y, z) + (t, 2t, 2t) = (y+z+t, y+2t, z+2t) \notin W_1 \cup W_3$$

12 Eigenvalues, eigenvectors, eigenspaces

Let f be a linear function

$$\begin{aligned}
f : V &\rightarrow V \\
\mathbb{R}^n &\rightarrow \mathbb{R}^n
\end{aligned}$$

and let $A \in \mathbb{R}^{n \times n}$ be

$$f(\vec{v}) = \vec{w}$$

the associated matrix

$$A\vec{v} = \vec{w}$$

Definition

We say that a non-zero vector $\vec{v} \in \mathbb{R}^n$ is an eigenvector of f (or equivalently an eigenvector of A) if there exist $\lambda \in \mathbb{R}$ such that

$$f(\vec{v}) = \lambda\vec{v} \Leftrightarrow A\vec{v} = \lambda\vec{v}$$

and λ is called the eigenvalue of the eigenvector \vec{v} .

Remark

Eigenvectors are special vectors of \mathbb{R}^n that are stretched or shrunked by the map f , that is \vec{v} and $f(\vec{v})$ are on the same line through the origin.

By definition, \vec{v} is an eigenvector those eigenvalue is λ

$$\Leftrightarrow f(\vec{v}) = \lambda\vec{v} \Leftrightarrow A\vec{v} = \lambda\vec{v} \Leftrightarrow A\vec{v} - \lambda\vec{v} = \vec{0}$$

$$\Leftrightarrow (A - \lambda I_n) \vec{v} = \vec{0} \Leftrightarrow B \vec{v} = \vec{0}$$

$$\lambda I_n \vec{v} = \lambda \vec{v}$$

$$A - \lambda I_n = B \in \mathbb{R}^{n \times n}$$

\vec{v} is an eigenvector with eigenvalue $\lambda \Leftrightarrow B\vec{v} = \vec{0}$

1. $B\vec{v} = \vec{0}$ has one and only one solution $\Leftrightarrow \text{rank}(B) = \text{rank}(B|\vec{0}) = n$

2. $B\vec{v} = \vec{0}$ has inf. many solutions $\Leftrightarrow \text{rank}(B) \equiv \text{rank}(B|\vec{0}) < n$

In case 1. we don't have any eigenvectors and λ is not an eigenvalue. In case 2. we have eigenvectors with eigenvalue λ .

$\text{rank}(B) = n$ number of pivots when B is in REF $\det(B) \neq 0$. $\text{rank}(B) < n$, then $\det(B) = 0$

Example

If

$$B = \begin{bmatrix} 1 & 2 & -1 \\ 0 & 5 & 1 \\ 0 & 0 & 3 \end{bmatrix} \quad \text{REF} \quad \text{rank}(B) = 3 = n$$

$$\det(B) = 1 \cdot 5 \cdot 3 \neq 0$$

If

$$B = \begin{bmatrix} 1 & 2 & -1 \\ 0 & 5 & 1 \\ 0 & 0 & 0 \end{bmatrix} \quad \text{REF} \quad \text{rank}(B) = 2 < n$$

$$\det(B) = 0$$

This λ is an eigenvalue if and only if

$$\det(B) = \det(A - \lambda I_n) = 0$$

and once we found there eigenvalues $\lambda \in \mathbb{R}$ then we solve the linear system

$$B\vec{v} = \vec{0}$$

where the unknonws are (v_1, \dots, v_n) in order to find the eigenvectos.

For instance from $\det(A - \lambda I_n) = 0$ where λ is unknonwn, we find some eigenvalues, let say $\lambda_1, \lambda_2, \dots, \lambda_m \in \mathbb{R}$ and for each eigenvalue we will find infinitely many eigenvectors solving

$$(A - \lambda_1 I_n)\vec{v} = \vec{0}$$

$$(A - \lambda_2 I_n)\vec{v} = \vec{0}$$

⋮

$$(A - \lambda_m I_n)\vec{v} = \vec{0}$$

12.1 Summary

Given a linear function $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$ whose associated matrix is $A \in \mathbb{R}^{n \times n}$ then

1. compute $\det(A - \lambda I_n) = p(\lambda)$ ¹ where λ is unknonwn.
2. Solve $p(\lambda) = 0$ and find the solutions (also called the roots of $p(\lambda)$) $\lambda_1, \dots, \lambda_m \in \mathbb{R}$ ($m \leq n$) which are the eigenvalues of f (of A).
3. For each eigenvalue λ_i (for $i = 1, \dots, m$) solve the linear system

$$(A - \lambda_i I_n)\vec{x} = \vec{0} \quad \text{homo. linear system}$$

¹this is a polynomial in λ of degree n and is called charateristic polynomial of A

whose unknonws are (x_1, \dots, x_n) . The non-zero solutions of this linear system are the eigenvector with eigenvalue λ_i

Definition

Given a linear function $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$, let λ be an eigenvalue of the set

$$\begin{aligned} E_\lambda &= \{\text{set of eigenvetors with eigenvalue } \lambda\} \cup \{\vec{0}\} \\ &= \left\{ \text{set of solutions of } (A - \lambda I_n) \vec{x} = \vec{0} \right\} \end{aligned}$$

is called the eigenspace of the eigenvalue λ .

Remark

An eigenspace is a vector space, because it is the set of solutions of a homogeneous linear system!

Let \vec{v}_1, \vec{v}_2 be solutions of a homogeneous linear system $C\vec{x} = 0$, thus $C\vec{v}_1 = \vec{0}, C\vec{v}_2 = \vec{0}$
If \vec{v}_1, \vec{v}_2 solutions, $\vec{v}_1 + \vec{v}_2$ is a solution?

$$C(\vec{v}_1 + \vec{v}_2) = C\vec{v}_1 + C\vec{v}_2 = \vec{0}$$

$$\begin{cases} x_1 + x_2 + x_3 = 0 \\ x_1 + x_2 = 0 \end{cases}$$