

Algèbre 02

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1 Vector Spaces

When you read the word vector you may immediately think of two points in \mathbb{R}^2 (or \mathbb{R}^3) connected by an arrow. Mathematically speaking, a vector is just an element of a vector space. This then begs the question : What is a vector space? Roughly speaking, a vector space is a set of objects that can be added and multiplied by scalars.

Definition 1.0.1 *A vector space is a set E of objects, called vectors, on which two operations called addition and scalar multiplication have been defined satisfying the following properties.*

If u, v, w are in E and if $\alpha, \beta \in \mathbb{R}$ are scalars :

1. *The sum $u + v$ is in E . (**closure under addition**)*
2. *$u + v = v + u$ (addition is commutative)*
3. *$(u + v) + w = u + (v + w)$ (addition is associative)*
4. *There is a vector in E called the zero vector, denoted by 0 , satisfying $v + 0 = v$.*
5. *For each v there is a vector $-v$ in E such that $v + (-v) = 0$.*
6. *The scalar multiple of v by α , denoted $\alpha \cdot v$, is in E . (**closure under scalar multiplication**)*
7. *$\alpha \cdot (u + v) = \alpha \cdot u + \alpha \cdot v$.*
8. *$(\alpha + \beta) \cdot v = \alpha \cdot v + \beta \cdot v$.*
9. *$(\alpha\beta) \cdot v = \alpha \cdot (\beta \cdot v)$.*
10. *$1 \cdot v = v$*

Remark 1.0.1 1. *Elements of E are called vectors, and elements of \mathbb{R} are called scalars. Instead of vector space on \mathbb{R} we also say, \mathbb{R} - vector space.*

2. *It can be shown that $0 \cdot v = 0$ for any vector v in E .*

To better understand the definition of a vector space, we first consider a few elementary examples.

Example 1.0.1 1. $\mathbb{R}^2, \mathbb{R}^3$ and more generally \mathbb{R}^n are real vector spaces.

2. *The set of applications from \mathbb{R} into \mathbb{R} is a vector space on \mathbb{R} .*

3. Let E be the unit disc in \mathbb{R}^2 :

$$E = \{(x, y) \in \mathbb{R}^2 / x^2 + y^2 \leq 1\}$$

The circle is not closed under scalar multiplication. For example, take $u = (1, 0) \in E$ and multiply by say $\alpha = 2$. Then $\alpha u = (2, 0)$ is not in E . Therefore, property (6) of the definition of a vector space fails, and consequently the unit disc is not a vector space.

4. Let E be the graph of the quadratic function $f(x) = x^2$:

$$E = \{(x, y) \in \mathbb{R}^2 / y = x^2\}$$

The set E is not closed under scalar multiplication. For example, $u = (1, 1)$ is a point in E but $2u = (2, 2)$ is not. You may also notice that E is not closed under addition either. For example, both $u = (1, 1)$ and $v = (2, 4)$ are in E but $u + v = (3, 5)$ and $(3, 5)$ is not a point on the parabola E . Therefore, the graph of $f(x) = x^2$ is not a vector space.

5. $\mathcal{F}(\mathbb{R}, \mathbb{R})$: **The vector space of functions from \mathbb{R} into \mathbb{R} .**

a/ Let f and g two elements of $\mathcal{F}(\mathbb{R}, \mathbb{R})$. The function $f + g$ is defined by :

$$\forall x \in \mathbb{R}, \quad (f + g)(x) = f(x) + g(x)$$

b/ If λ is a real number and f is a function of $\mathcal{F}(\mathbb{R}, \mathbb{R})$, the function $\lambda.f$ is defined by the image of any real x as follows :

$$\forall x \in \mathbb{R}, \quad (\lambda.f)(x) = \lambda f(x)$$

c/ **The identity** The identity for addition is the null function, defined by :

$$\forall x \in \mathbb{R}, \quad f(x) = 0.$$

This function can be written $0_E = 0_{\mathcal{F}(\mathbb{R}, \mathbb{R})}$.

d/ **The inverses** The inverse of f in $\mathcal{F}(\mathbb{R}, \mathbb{R})$ is the function g from \mathbb{R} to \mathbb{R} defined by :

$$\forall x \in \mathbb{R}, \quad g(x) = -f(x).$$

The inverse of f is noted $-f$.

6. Let $E = \mathbb{R}_2[X] = \{P = aX^2 + bX + c, \quad a, b, c \in \mathbb{R}\}$ be the set of polynomials of degree less than or equal to 2, with coefficients in \mathbb{R} , provided with the following operations :

a/ A law " + ", given by : $\forall P, Q \in E, \quad P = aX^2 + bX + c, \quad Q = a'X^2 + b'X + c'$,

$$P + Q = (a + a')X^2 + (b + b')X + (c + c').$$

b/ A law " \cdot " defined by : $\forall \alpha \in \mathbb{R}, \quad \forall P \in E, \quad P = aX^2 + bX + c,$

$$\alpha \cdot P = (\alpha a)X^2 + (\alpha b)X + (\alpha c).$$

$(E, +, \cdot)$ is a vectorial space on \mathbb{R} .

1.1 Subspaces of Vector Spaces

Frequently, one encounters a vector space F that is a subset of a larger vector space E . In this case, we would say that F is a subspace of E . Below is the formal definition.

Definition 1.1.1 *Let E be a vector space. A subset F of E is called a subspace of E if it satisfies the following properties :*

1. *The zero vector of E is also in F .*
2. *F is closed under addition, that is, if u and v are in F then $u + v$ is in F .*
3. *F is closed under scalar multiplication, that is, if u is in F and α is a scalar then $\alpha \cdot u$ is in F .*

Example 1.1.1 *Let F be the graph of the function $f(x) = 2x$:*

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

F a subspace of $E = \mathbb{R}^2$.

If $x = 0$ then $y = 2 \cdot 0 = 0$ and therefore $(0, 0)$ is in F .

Let $u = (a, 2a)$ and $v = (b, 2b)$ be elements of F . Then $u + v = (a, 2a) + (b, 2b) = (a + b, 2a + 2b) = (a + b, 2(a + b))$. Because the x and y components of $u + v$ satisfy $y = 2x$ then $u + v$ is inside in F . Thus, F is closed under addition.

Let α be any scalar and let $u = (a, 2a)$ be an element of F . Then $\alpha u = (\alpha a, \alpha 2a) = (\alpha a, 2\alpha a) \in F$. F is closed under scalar multiplication.

All three conditions of a subspace are satisfied for F and therefore F is a subspace of E .

Example 1.1.2 *Let F be the first quadrant in \mathbb{R}^2 :*

$$F = \{(x, y) \in \mathbb{R}^2 \mid x \geq 0, y \geq 0\}.$$

The set F contains the zero vector and the sum of two vectors in F is again in F . However, F is not closed under scalar multiplication. For example if $u = (1, 1)$ and $\alpha = -1$, then $\alpha u = (-1, -1)$ is not in F because the components of αu are clearly not non-negative.

Example 1.1.3 *Let $E = \mathbb{R}_n[t]$ and consider the subset F of E :*

$$F = \{P(t) \in \mathbb{R}_n[t] \mid P'(1) = 0\}$$

F is a subspace of E .

The zero polynomial $0(t)$ clearly has derivative at $t = 1$ equal to zero, that is, $0'(1) = 0$, and thus the zero polynomial is in F . Now suppose that $P(t)$ and $Q(t)$ are two polynomials in F . Then, $P'(1) = 0$ and also $Q'(1) = 0$, from the rules of differentiation, we compute $(P + Q)'(1) = P'(1) + Q'(1) = 0 + 0$.

Therefore, the polynomial $(P + Q)(t)$ is in F , and thus F is closed under addition.

Now let α be any scalar and let $P(t)$ be a polynomial in F . Then $P'(1) = 0$. Using the rules of differentiation, we compute that $(\alpha P)'(1) = \alpha P'(1) = \alpha \cdot 0 = 0$. Therefore, the polynomial $(\alpha P)(t)$ is in F and thus F is closed under scalar multiplication.

All three properties of a subspace hold for F and therefore F is a subspace of $\mathbb{R}_n[t]$.

Example 1.1.4 1. *Any field \mathbb{K} is a vectorspace on \mathbb{K} .*

2. Any field \mathbb{L} containing a field \mathbb{K} is a vector space on \mathbb{K} and \mathbb{K} is a vector subspace of \mathbb{L} .
3. \mathbb{C} is a vector space on \mathbb{R} and \mathbb{R} is a subspace of \mathbb{C} .

Example 1.1.5 Consider $F = \{(x, y) \in \mathbb{R}^2 / x^2 + y^2 < 0\}$, $F = \emptyset$, so F is not a subspace of \mathbb{R}^2 .

Example 1.1.6 Let $F = \{(x, y) \in \mathbb{R}^2 / x - y + 1 = 0\}$, we have $0_{\mathbb{R}^2} = (0, 0) \notin F$, since $0 - 0 + 1 \neq 0$ therefore F is not a subspace of \mathbb{R}^2 .

Example 1.1.7 Let $F = \{(x, y) \in \mathbb{R}^2 / xy \geq 0\}$, we have $(2, 1), (-1, -2) \in F$, but $(2, 1) + (-1, -2) = (1, -1) \notin F$ because does not check $xy \geq 0$ so F is not a subspace of \mathbb{R}^2 .

1.2 Operation on vector subspaces

Proposition 1.2.1 Let \mathbb{K} be a field, E a \mathbb{K} -vector space, F and G two subspaces of E , then :

1. $F \cap G$ is a subspace of E .
2. $F \cup G$ is a subspace of E if and only if, $F \subset G$ or $G \subset F$.

Proof 1.2.1 (of 1.) We have F and G are subspaces of E , then : ($F \subset E$ and $G \subset E$ therefore $F \cap G \subset E$.)

a/ $0_E \in F$ and $0_E \in G$ which means that $0_E \in F \cap G$.

b/ $\forall \alpha, \beta \in \mathbb{K}, \forall x, y \in F \cap G$ (i.e. $x \in F \wedge x \in G$), we have $\alpha x + \beta y \in F$ and $\alpha x + \beta y \in G$, therefore $\alpha x + \beta y \in F \cap G$. Then $F \cap G$ is a subspace of E .

Remark 1.2.1 We generalize the property (1) to any family of vector subspaces, i.e. If $(F_i)_{i \in I, I \subset \mathbb{N}}$, is a family of subvector spaces, then $\bigcap_{i \in I} F_i$ is a subspace.

Example 1.2.1 Let $E = \mathbb{R}^2$ be the vector space on \mathbb{R} . Consider the following subspaces F and G :

$$F = \{(x, y) \in \mathbb{R}^2 / y = 0\}, \quad G = \{(x, y) \in \mathbb{R}^2 / x = 0\}.$$

F and G are the x -axis and y -axis respectively.

Since $(1, 0) \in F$ with $(1, 0) \notin G$, then $F \not\subset G$ and $(0, 1) \in G$ with $(0, 1) \notin F$, then $G \not\subset F$. Therefore, $F \cup G$ is not a subspace of \mathbb{R}^2 .

The result can be obtained by noting that $(1, 0), (0, 1) \in F \cup G$ but $(1, 0) + (0, 1) = (1, 1) \notin F$ and $(1, 1) \notin G$ then $(1, 1) \notin F \cup G$. This means that $F \cup G$ is not a subspace of E .

Theoreme 1.2.1 Let \mathbb{K} be a field, E a vector space on \mathbb{K} , F and G two subspaces of E . The set $F + G$ defined by

$$F + G = \{x + y / x \in F \text{ and } y \in G\} \subset E$$

is a subspace of E called **sum** of the subspaces F and G . If in addition $F \cap G = \{0_E\}$, we say that the sum $F + G$ is a **direct sum** and we write $F \oplus G$.

Proof 1.2.2 $F + G$ is a subspace of E :

1. $0_E = 0_E + 0_E \in F + G$ because $0_E \in F$ and $0_E \in G$ since F and G are two subspaces of E .

2. $\forall \alpha, \beta \in \mathbb{K}, \quad \forall z, z' \in F + G$, then $z = x + y$ and $z' = x' + y'$ with $x, x' \in F$ and $y, y' \in G$.
Since F and G are subspaces of E , then

$$\alpha x + \beta x' \in F \quad \text{and} \quad \alpha y + \beta y' \in G.$$

This means that $(\alpha x + \beta x') + (\alpha y + \beta y') \in F + G$.

Therefore $(\alpha x + \beta x') + (\alpha y + \beta y') = \alpha(x + y) + \beta(x' + y') \in F + G$, i.e. $\alpha z + \beta z' \in F + G$

Example 1.2.2 Consider the vector space \mathbb{R}^3 , the subspaces F and H given by

$$F = \{(x, y, z) \in \mathbb{R}^3 / x + y - z = 0\} \quad \text{and} \quad H = \{(x, y, z) \in \mathbb{R}^3 / x = y = 0\}.$$

We have $F + G = F \oplus G$. Indeed :

Let $(x, y, z) \in F \cap H$, so $(x, y, z) \in F$, i.e. $z = x + y$ and $(x, y, z) \in H$ i.e. $x = y = 0$, so $x = y = z = 0$, therefore $F \cap H = \{0_{\mathbb{R}^3}\}$.

Example 1.2.3 For any vector space E , there are two trivial subspaces in E , namely, E itself is a subspace of E and the set consisting of the zero vector $F = \{0\}$ is a subspace of E .

There is one particular way to generate a subspace of any given vector space E using the span of a set of vectors.

2 Linear combinations, generating families, linearly independent families, bases, dimension.

2.1 Linear combinations

Let v_1, v_2, \dots, v_n be a family of vectors of a vector space on \mathbb{K} , We call linear combination of these vectors any vector of type

$$v = \lambda_1 v_1 + \lambda_2 v_2 + \dots + \lambda_n v_n.$$

The scalars $\lambda_1, \dots, \lambda_n$ are called the coefficients of the linear combination.

The span of $\{v_1, v_2, \dots, v_n\}$ is the set of all linear combinations of v_1, v_2, \dots, v_n .

$$\text{span} \{v_1, v_2, \dots, v_n\} = \{\lambda_1 v_1 + \lambda_2 v_2 + \dots + \lambda_n v_n / \lambda_1, \dots, \lambda_n \in \mathbb{K}\}$$

The span of a set of vectors in E is a subspace of E .

2.2 Generating families

Definition 2.2.1 The family $\{v_1, v_2, \dots, v_n\}$ is a generating family of the vector space E if every vector of E is a linear combination of the vectors v_1, v_2, \dots, v_n . This can also be written :

$$\forall v \in E, \quad \exists \lambda_1, \lambda_2, \dots, \lambda_n \in \mathbb{K} / v = \lambda_1 v_1 + \lambda_2 v_2 + \dots + \lambda_n v_n$$

We also say that the family $\{v_1, v_2, \dots, v_n\}$ generates the vector space E and we write

$$E = \text{span} \{v_1, v_2, \dots, v_n\}$$

.

Example 2.2.1 Let the vectors $v_1 = (2, 1), v_2 = (1, 1) \in \mathbb{R}^2$. The vectors $\{v_1, v_2\}$ form a generating family of \mathbb{R}^2 . Indeed, let $v = (x, y) \in \mathbb{R}^2$, showing that v is a linear combination of v_1 and v_2 is equivalent to demonstrate the existence of two real numbers α and β such that $v = \alpha v_1 + \beta v_2$. So we need to study the existence of solutions to the system :

$$\begin{aligned} 2\alpha + \beta &= x \\ \alpha + \beta &= y \end{aligned}$$

Its solutions are $\alpha = x - y$ and $\beta = -x + 2y$, whatever the real numbers x and y . This proves that there can be several different finite families, not included in each other, generating the same vector space.

Example 2.2.2 Let $E = \mathbb{R}_n[X]$ be the vector space of polynomials of degree $\leq n$. Then the polynomials $\{1, X, \dots, X^n\}$ form a generating family of E .

2.3 Linearly independent families

Definition 2.3.1 1. A family $\{v_1, v_2, \dots, v_n\}$ of vectors of a vector space E is linearly independent if the only linear combination of these vectors equal to the zero vector is the one whose coefficients are all zero. We also say that vectors $\{v_1, v_2, \dots, v_n\}$ are linearly independent.

This can be expressed as :

$\{v_1, v_2, \dots, v_n\}$ is a linearly independent family is equivalent to :

$$((\lambda_1, \dots, \lambda_n) \in \mathbb{K}^n \text{ and } \lambda_1 v_1 + \lambda_2 v_2 + \dots + \lambda_n v_p = 0_E) \Rightarrow \lambda_1 = \lambda_2 = \dots = \lambda_n = 0.$$

2.4 Linearly dependent families

Definition 2.4.1 1. A non linearly independent family is called a linearly dependent family. We also say that vectors $\{v_1, v_2, \dots, v_n\}$ are linearly dependents.

This can be expressed as : $\{v_1, v_2, \dots, v_n\}$ is a linearly dependent family is equivalent to

$$(\exists (\lambda_1, \dots, \lambda_n) \in \mathbb{K}^n - \{0_{\mathbb{K}^n}\} / \lambda_1 v_1 + \lambda_2 v_2 + \dots + \lambda_n v_p = 0_E).$$

Example 2.4.1 The polynomials $P_1(X) = 1 - X, P_2(X) = 5 + 3X - 2X^2$ and $P_3(X) = 1 + 3X - X^2$ form a linearly dependent family in the vector space $\mathbb{R}_2[X]$, because $3P_1(X) - P_2(X) + 2P_3(X) = 0$.

Example 2.4.2 In the vector space $\mathcal{F}(\mathbb{R}, \mathbb{R})$ of functions from \mathbb{R} into \mathbb{R} , consider the family $\{\cos, \sin\}$. Let's show that it's a linearly independent family.

Suppose we have $\lambda \cos + \mu \sin = 0$, which is equivalent to $\forall x \in \mathbb{R}, \lambda \cos(x) + \mu \sin(x) = 0$. In particular, for $x = 0$, this equality gives $\lambda = 0$. And for $x = \pi/2$, it gives $\mu = 0$. So $\{\cos, \sin\}$ is a linearly independent family.

On the other hand, the family $\{\cos^2, \sin^2, 1\}$ is linearly dependent because we have : $\cos^2 + \sin^2 - 1 = 0$.

The coefficients of the linear dependence are $\lambda_1 = 1, \lambda_2 = 1, \lambda_3 = -1$.

Example 2.4.3 In the vector space \mathbb{R}^4 defined over the field \mathbb{R} , consider the following vectors :

$$v_1 = (1, 0, -1, 1), v_2 = (0, 1, 1, 0), v_3 = (1, 0, 0, 1), v_4 = (0, 0, 0, 1), v_5 = (1, 1, 0, 1).$$

The set $\{v_1, v_2, v_3, v_4\}$ is linearly independent (to be verified). The set $S_2 = \{v_1, v_2, v_5\}$ is linearly dependent ($v_5 = v_1 + v_2$).

Theoreme 2.4.1 Let E be a vector space over the field \mathbb{K} . A set $F = \{v_1, v_2, \dots, v_n\}$ of n vectors of E , ($n > 2$) is linearly dependent if and only if at least one of the vectors of F is a linear combination of the other vectors of F .

- Remark 2.4.1**
1. Any family containing a linearly dependent family is linearly dependent.
 2. Any family included in a linearly independent family is linearly independent.
 3. $\{v\}$ is linearly independent if and only if $v \neq 0$.
 4. Any set containing the null vector is linearly dependent.

2.5 Basis

A basis of a vector space is linearly independent generating family.

If $B = (x_i)_{i \in I}$, $I \subset \mathbb{N}$ is a basis of E , then any $x \in E$ is uniquely written as a linear combination of elements of B .

$$x = \sum_{i \in I} \alpha_i x_i$$

The scalars $(\alpha_i)_{i \in I}$, are called the coordinates of x in the basis B .

3 Finite dimensional vector spaces

Definition 3.0.1 If a vector space is spanned by a finite number of vectors, it is said to be **finite-dimensional**.

Otherwise it is **infinite-dimensional**. The number of vectors in a basis for a finite-dimensional vector space E is called the **dimension** of E and denoted $\dim E$.

By convention, we say that $\{0_E\}$ is a finite-dimensional space.

Definition 3.0.2 A family $\{v_1, \dots, v_n\}$ of vectors of E is said to be a basis of E if and only if, we have :

1. $\{v_1, \dots, v_n\}$ is a linearly independent family of E and
2. $\{v_1, \dots, v_n\}$ is a generating family of E .

Example 3.0.1 1. The set $(1, i)$ is a basis of the \mathbb{R} -vector space \mathbb{C} .

Indeed, if $a, b \in \mathbb{R}$ are such that $a \cdot 1 + b \cdot i = 0$ then $a + ib = 0 + i0$ and therefore $a = b = 0$.

The set is therefore linearly independent.

For any complex number, there are $a, b \in \mathbb{R}$ such that $z = a + ib$, then $(1, i)$ is a generating set of \mathbb{C} , it is therefore a basis of \mathbb{C} .

2. In \mathbb{R}^3 , the set $\{e_1 = (1, 0, 0), e_2 = (0, 1, 0), e_3 = (0, 0, 1)\}$ forms a basis of \mathbb{R}^3 , called **canonical basis** of \mathbb{R}^3 .

The set $\{v_1 = (1, 0, 1), v_2 = (1, -1, 1), v_3 = (0, 1, 1)\}$ is a basis of \mathbb{R}^3 . Indeed :

a/ The family is linearly independent.

Let $\alpha_1, \alpha_2, \alpha_3 \in \mathbb{R}$ such that $\alpha_1 v_1 + \alpha_2 v_2 + \alpha_3 v_3 = 0_{\mathbb{R}^3}$. Then

$$\begin{aligned} \alpha_1 + \alpha_2 &= 0 \\ \alpha_2 + \alpha_3 &= 0 \\ \alpha_1 + \alpha_2 + \alpha_3 &= 0 \end{aligned}$$

which leads to $\alpha_1 = \alpha_2 = \alpha_3 = 0$.

b/ The set is generating of \mathbb{R}^3 . Let $(x, y, z) \in \mathbb{R}^3$. We are looking for $\alpha_1, \alpha_2, \alpha_3 \in \mathbb{R}$ such that $(x, y, z) = \alpha_1 v_1 + \alpha_2 v_2 + \alpha_3 v_3$. We then obtain the system

$$\begin{aligned}\alpha_1 + \alpha_2 &= x \\ \alpha_2 + \alpha_3 &= y \\ \alpha_1 + \alpha_2 + \alpha_3 &= z\end{aligned}$$

and we find $\alpha_1 = 2x + y - z$, $\alpha_2 = x - y + z$ and $\alpha_3 = -x + z$.

So $\text{span}\{v_1 = (1, 0, 1), v_2 = (1, -1, 1), v_3 = (0, 1, 1)\} = \mathbb{R}^3$. Then $\{v_1, v_2, v_3\}$ is a basis of \mathbb{R}^3 .

More generally, we have :

Proposition 3.0.1 Canonical base of \mathbb{K}^n

Consider the vector space $E = \mathbb{K}^n$ over the field \mathbb{K} .

The standard basis vectors of E are a specific set of basis vectors that are commonly used in linear algebra. They are the unit vectors in each dimension of the vector space :

(e_1, e_2, \dots, e_n) of \mathbb{K}^n called **canonical** and given by :

$$e_1 = (1, 0, 0, \dots, 0), e_2 = (0, 1, 0, 0, \dots, 0), \dots, e_n = (0, 0, 0, \dots, 0, 1).$$

Proposition 3.0.2 Canonical base of $\mathbb{K}_n[X]$

Let $n \in \mathbb{N}$. Consider the vector space $E = \mathbb{K}_n[X]$ of polynomials of degree $\leq n$ with coefficients in \mathbb{K} . There is a specific basis of $\mathbb{K}_n[X]$ called **canonical**, given by $\{1, X, X^2, \dots, X^n\}$.

Theoreme 3.0.1 Theorem of the extracted basis From any finite generating family of E , we can extract a basis of E . In particular, a finite-dimensional space admits a basis.

Theoreme 3.0.2 Incomplete basis theorem If E is finite-dimensional, then any linearly independent family of E can be completed into a basis of E . To complete it, simply consider certain vectors of a generating family of E .

Theoreme 3.0.3 Dimension If E is finite-dimensional, then all bases of E have the same number of vectors (dimension of E).

Corollary 3.0.1 If E is a finite-dimensional vector space ($\dim E = n$) and if $B = (v_1, v_2, \dots, v_n)$ is a family of n vectors of E , then the following conditions are equivalent :

1. B is linearly independent.
2. B is a generating set of E .
3. B is a basis of E .

Remark 3.0.1 1. In particular, in a n -dimensional space, a linearly independent set always has at most n elements, and a generating family always has at least n elements.

2. If E and F are finite-dimensional, then $\dim(E \times F) = \dim(E) + \dim(F)$. In particular, $\dim(\mathbb{K}^n) = n$.
3. $\dim(\mathbb{K}_n[X]) = n + 1$.

Definition 3.0.3 If (v_1, v_2, \dots, v_n) is a finite set of E , we call **rank** of (v_1, v_2, \dots, v_n) the dimension of $F = \text{Vect}(v_1, v_2, \dots, v_n)$.

Let $G = \{v_1 = (2, 1), v_2 = (4, 2), v_3 = (-3, 4)\}$ be a subset of \mathbb{R}^2 . Let's determine the rank of G .

The set G is linearly dependent ($v_2 = 2v_1$), so $\text{span}(v_1, v_2, v_3) = \text{span}(v_2, v_3)$, so $\text{rank}(G) = 2$.

3.1 Subspaces and dimension

If E is a finite-dimensional vector space and if F is a subspace of E , then we have $\dim(F) \leq \dim(E)$ and Furthermore :

$$\dim(F) = \dim(E) \Leftrightarrow F = E.$$

Grassmann formula : Let E be a finite-dimensional vector space and let F, G be two subspaces of E . Then

$$\dim(F + G) = \dim(F) + \dim(G) - \dim(F \cap G).$$

In particular, F and G are in direct sum if and only if

$$\dim(F + G) = \dim(F) + \dim(G).$$