

# MAPS BETWEEN SETS

Let  $A$ ,  $B$  and  $C$  be nonempty sets.

**Definition** A **map**  $f : A \rightarrow B$  is a correspondence assigning to each element of  $A$  a unique element in  $B$ . The set  $A$  is called **domain** of the map  $f$  and the set  $B$  is the **range** of  $f$ .

Given  $a \in A$  we denote by  $f(a)$  the image element of  $a$  in  $B$  through the map  $f$ . The **image** of  $f$  is the set

$$\text{Im}(f) = f(A) = \{f(a) \mid a \in A\}.$$

The **inverse image** of an element  $b \in B$  is the set

$$f^{-1}(b) = \{a \in A \mid f(a) = b\}.$$

The **identity map**  $A$  is the map  $id_A : A \rightarrow A$  defined by  $id_A(a) = a$  for every  $a \in A$ .

**Definition** Given a map  $f : A \rightarrow B$  we call  $f$ :

1. **Injective or one-to-one** if given two different elements of  $A$  their images through  $f$  are distinct. That is

$$\forall a_1, a_2 \in A, a_1 \neq a_2 \Rightarrow f(a_1) \neq f(a_2).$$

Equivalently,

$$\text{if } \exists a_1, a_2 \in A \text{ such that } f(a_1) = f(a_2) \Rightarrow a_1 = a_2.$$

2. **Surjective** if every element in  $B$  is the image through  $f$  of an element of  $A$ . That is,  $f(A) = B$ ,

$$\forall b \in B, \exists a \in A \text{ such that } f(a) = b.$$

3. **Bijjective** if it is injective and surjective.

**Definition** Given two maps  $f : A \rightarrow B$  and  $g : B \rightarrow C$ , the **composition** of  $f$  with  $g$  is the map  $g \circ f : A \rightarrow C$  defined by  $(g \circ f)(a) = g(f(a))$  for every  $a \in A$ .

**Definition** Given a bijective map  $f : A \rightarrow B$  the **inverse** map of  $f$  is the map  $f^{-1} : B \rightarrow A$  defined by:

$$\forall b \in B, f^{-1}(b) = a \text{ if } f(a) = b \text{ with } a \in A.$$

It holds that  $f^{-1} \circ f = id_A$  and that  $f \circ f^{-1} = id_B$ .

## 7. LINEAR TRANSFORMATIONS

Let  $V$  and  $W$  be real vector spaces.

**Definition** A map  $f : V \rightarrow W$  is a **linear transformation** (or homomorphism) if it verifies

$$\begin{aligned}\forall u, v \in V, f(u + v) &= f(u) + f(v), \\ \forall \lambda \in \mathbb{R}, f(\lambda u) &= \lambda f(u).\end{aligned}$$

This is equivalent to

$$\forall u, v \in V, \forall \lambda, \mu \in \mathbb{R}, f(\lambda u + \mu v) = \lambda f(u) + \mu f(v).$$

A linear transformation  $f : V \rightarrow W$  where  $W = V$  is called **endomorphism**.

**Example** The map  $f : \mathbb{R}^3 \longrightarrow \mathbb{R}^3$  defined by

$$f(x, y, z) = (2x - y + 4z, 3x - z, 6x + y)$$

is linear.

It is also linear every transformation  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  defined by

$$f(x_1, \dots, x_n) = \left( \sum_{i=1}^n a_i x_i, \sum_{i=1}^n b_i x_i, \dots, \sum_{i=1}^n l_i x_i \right),$$

with  $a_i, b_i, \dots, l_i \in \mathbb{R}$ ,  $i = 1, \dots, n$ .

**Proposition** Let  $f : V \rightarrow W$  be a linear transformation. Given a set  $\{u_1, \dots, u_m\}$  of vectors in  $V$ , the following statements hold:

1. Given  $\lambda_1, \dots, \lambda_m \in \mathbb{R}$  then

$$f\left(\sum_{i=1}^m \lambda_i u_i\right) = \sum_{i=1}^m \lambda_i f(u_i).$$

In particular,  $f(0_V) = 0_W$  and  $f(-u) = -f(u)$ ,  $\forall u \in V$ .

2. If  $u_1, \dots, u_m$  are linearly dependent then

$$f(u_1), \dots, f(u_m)$$

are also linearly dependent.



3. If  $f(u_1), \dots, f(u_m)$  are linearly independent then  $u_1, \dots, u_m$  are linearly independent. The converse is not true, in general, the linear independence of vectors is not preserved by linear transformations.
4. If  $U$  is a vector subspace of  $V$  with basis  $\{u_1, \dots, u_m\}$  then  $f(U) = \langle f(u_1), \dots, f(u_m) \rangle$ . This is  $\{f(u_1), \dots, f(u_m)\}$  is a generating set of  $f(U)$  but it may not be a basis of  $f(U)$ .

**Examples** Let us consider the linear transformation  $f : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  defined by

$$f(x, y, z) = (x, y).$$

1. The vectors  $v_1 = (1, 0, 0)$  and  $v_2 = (0, 1, 0)$  are linearly independent and so are their images,  $f(v_1) = (1, 0)$  and  $f(v_2) = (0, 1)$ . On the other hand, the vectors  $u_1 = (1, 0, 1)$  and  $u_2 = (2, 0, 0)$  are linearly independent but their images  $f(u_1) = (1, 0)$  and  $f(u_2) = (2, 0)$  are linearly dependent.
2. Given the subspace  $U = \langle u_1 = (1, 0, 1), u_2 = (2, 0, 0) \rangle$ , with  $\dim U = 2$ , of  $\mathbb{R}^3$  then  $f(U) = \langle f(u_1), f(u_2) \rangle = \langle (1, 0) \rangle$  and so  $\dim f(U) = 1$ .

## THE MATRIX OF A LINEAR TRANSFORMATION

**Proposition** Let  $B = \{v_1, \dots, v_n\}$  be a basis of  $V$  and  $\{w_1, \dots, w_n\}$  a set of vectors in  $W$ . There exists a unique linear transformation  $f : V \rightarrow W$  such that

$$f(v_1) = w_1, \dots, f(v_n) = w_n.$$

**Example** Let  $f : \mathbb{R}^3 \rightarrow \mathbb{R}^4$  be a linear application verifying:

$$f(1, 0, 0) = (3, 1, 2, 4), f(0, 1, 0) = (1, -1, -5, 5),$$

$$f(0, 0, 1) = (2, -2, -3, 4).$$

By the previous proposition, there exists a unique linear transformation verifying the previous conditions. Such transformation is defined by:

$$\begin{aligned} f(x, y, z) &= xf(1, 0, 0) + yf(0, 1, 0) + zf(0, 0, 1) = \\ &= (3x + y + 2z, x - y - 2z, 2x - 5y - 3z, 4x + 5y + 4z). \end{aligned}$$

**Definition** Let  $B_V = \{v_1, \dots, v_n\}$  be a basis of  $V$  and let  $(x_1, \dots, x_n)$  be the coordinates of a generic vector in  $V$  with respect to the basis  $B_V$ . Let  $B_W = \{w_1, \dots, w_m\}$  be a basis of  $W$  and let  $(y_1, \dots, y_m)$  be the coordinates of a generic vector in  $W$  with respect to the basis  $B_W$ .

Let us suppose that  $f : V \rightarrow W$  is a linear transformation such that

$$f(v_j) = \sum_{i=1}^m a_{ij} w_i = a_{1j} w_1 + a_{2j} w_2 + \dots + a_{mj} w_m,$$

this is, the coordinates of  $f(v_j)$  in the basis  $B_W$  are  $(a_{1j}, \dots, a_{mj})$ .

The **matrix expression of  $f$**  in  $B_V$  of  $V$  and  $B_W$  of  $W$  is

$$\begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ & & \dots & \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}.$$

We say that  $A = (a_{ij})$  is the **matrix of the linear transformation  $f$**  with respect to the basis  $B_V$  of  $V$  and  $B_W$  of  $W$ , which is denoted by  $M_f(B_V, B_W)$ .

If  $f : V \rightarrow V$  is an endomorphism we denote by  $M_f(B_V)$  the matrix  $M_f(B_V, B_V)$ .

**Example** Let  $f : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be the linear transformation verifying

$$f(1, 1, 1) = (2, 2), f(0, 1, 1) = (1, 1), f(0, 0, 3) = (0, 3).$$

We obtain next the matrix expression of  $f$  in the standard basis  $B_3$  of  $\mathbb{R}^3$  and  $B_2$  of  $\mathbb{R}^2$ .

For this purpose we compute the images of the vectors in the basis  $B_3$ .

$$f(1, 0, 0) = f(1, 1, 1) - f(0, 1, 1) = (2, 2) - (1, 1) = (1, 1),$$

$$f(0, 1, 0) = f(0, 1, 1) - \frac{1}{3}f(0, 0, 3) = (1, 1) - (0, 1) = (1, 0),$$

$$f(0, 0, 1) = \frac{1}{3}f(0, 0, 3) = (0, 1).$$

The columns of  $M_f(B_3, B_2)$  are the coordinates in the basis  $B_2$  of such images. Thus, the matrix expression is

$$\begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}.$$

Then  $f(x_1, x_2, x_3) = (x_1 + x_2, x_1 + x_3)$  for every  $(x_1, x_2, x_3) \in \mathbb{R}^3$ .

## KERNEL AND IMAGE

**Proposition** Let  $f : V \longrightarrow W$  be a linear transformation. The following statements hold:

1. The image  $\text{Im}(f) = \{f(v) \mid v \in V\}$  is a vector subspace of  $W$ , which is called the **image** of the linear transformation  $f$ .
2. If  $\{v_1, \dots, v_n\}$  is a generating set of  $V$  then  $\{f(v_1), \dots, f(v_n)\}$  is a generating set of  $\text{Im}(f)$ . We call **rank** of  $f$  to the dimension of  $\text{Im}(f)$

$$\text{rank}(f) = \dim(\text{Im}(f)).$$

3. We define the **kernel** of the linear transformation  $f$  as the set of vectors in  $V$  whose image through  $f$  is the zero vector  $0_W$ ,

$$\text{Ker}(f) = \{v \in V \mid f(v) = 0_W\}.$$

Then  $\text{Ker}(f)$  is a vector subspace of  $V$  and  $\text{Ker}(f) = f^{-1}(0_W)$ .

**Proposition** Let  $f : V \rightarrow W$  be a linear transformation. If  $V$  is a finitely generated vector space then:

$$\dim(\text{Ker}(f)) + \dim(\text{Im}(f)) = \dim V.$$

**Proposition** The linear transformation  $f : V \rightarrow W$  is injective if and only if

$$\text{Ker}(f) = \{0_V\}.$$



**Proposition** Let  $f : V \rightarrow W$  be a linear transformation.

1. If  $V$  is finitely generated, then  $f$  is injective if and only if  $\dim V = \dim(f(V))$ .
2. Let  $B = \{v_1, \dots, v_n\}$  be a basis of  $V$ .  $f$  is one-to-one if and only if  $f(B) = \{f(v_1), \dots, f(v_n)\}$  is a basis of  $\text{Im}(f)$  if and only if  $f(B)$  is linearly independent.

**Definition** An **isomorphism** is a linear and bijective transformation  $f : V \rightarrow W$ . If  $f$  is an isomorphism then the vector spaces  $V$  and  $W$  are **isomorphic**.

## Proposition

1. A linear transformation  $f : V \rightarrow W$  is bijective if and only if

$$\text{Im}(f) = W \text{ and } \text{Ker}(f) = \{0_V\}.$$

2. An endomorphism  $f : V \rightarrow V$  is bijective if and only if  $f$  is injective and surjective.

**Example** Let  $h : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be the linear transformation defined by  $h(x_1, x_2, x_3) = (2x_1 + 2x_2 - 3x_3, 3x_1 + x_2 - 2x_3)$ . We obtain next the cartesian equations of  $\text{Ker}(h)$  and a basis of  $\text{Im}(h)$  in  $B_3$  of  $\mathbb{R}^3$  and  $B_2$  of  $\mathbb{R}^2$ .

The matrix of  $h$  is

$$M_h(B_3, B_2) = \begin{pmatrix} 2 & 2 & -3 \\ 3 & 1 & -2 \end{pmatrix}.$$

A vector  $(x_1, x_2, x_3)$  in  $\mathbb{R}^3$  belongs to  $\text{Ker}(h)$  if

$$\begin{pmatrix} 2 & 2 & -3 \\ 3 & 1 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

equivalently

$$\text{Cartesian equations of } \text{Ker}(h) \begin{cases} 2x_1 + 2x_2 - 3x_3 = 0 \\ 3x_1 + x_2 - 2x_3 = 0, \end{cases}$$

Also

$$\begin{aligned} \text{Im}(h) &= \langle h(1, 0, 0), h(0, 1, 0), h(0, 0, 1) \rangle = \\ &= \langle (2, 3), (2, 1), (-3, -2) \rangle = \langle (2, 3), (2, 1) \rangle. \end{aligned}$$

Then  $\{(2, 3), (2, 1)\}$  is a basis of  $\text{Im}(h)$ .

**Theorem** Let us suppose that  $\dim V = n$ , then  $V$  is isomorphic to  $\mathbb{R}^n$ .

**Example** Let  $V$  be a real vector space with  $\dim V = 3$ . Let  $B = \{u_1, u_2, u_3\}$  be a basis of  $V$  and let  $B_3 = \{e_1, e_2, e_3\}$  be the standard basis of  $\mathbb{R}^3$ . The linear transformation  $f : V \rightarrow \mathbb{R}^3$  determined by the conditions  $f(u_1) = e_1$ ,  $f(u_2) = e_2$  and  $f(u_3) = e_3$  is an isomorphism. Then  $V$  and  $\mathbb{R}^3$  are isomorphic.

## OPERATIONS WITH LINEAR TRANSFORMATIONS

Let  $B_V$  and  $B_W$  be basis of  $V$  and  $W$  respectively. Given two linear transformations  $f : V \rightarrow W$ ,  $g : V \rightarrow W$  and a scalar  $\lambda \in \mathbb{R}$ , we define the following operations between linear transformations:

1. **Sum**  $f + g : V \rightarrow W$  given by  $(f + g)(v) = f(v) + g(v)$  for every  $v \in V$ .
2. **Multiplication by a scalar**  $\lambda f : V \rightarrow W$  given by  $(\lambda f)(v) = \lambda f(v)$  for every  $v \in V$ .

It holds that

$$\begin{aligned}M_{f+g}(B_V, B_W) &= M_f(B_V, B_W) + M_g(B_V, B_W), \\M_{\lambda f}(B_V, B_W) &= \lambda M_f(B_V, B_W).\end{aligned}$$

Let  $U$  be a real vector space and  $B_U$  a basis of  $U$ . Given two linear transformations  $f : V \rightarrow W$  and  $g : W \rightarrow U$  the **composition**  $g \circ f : V \rightarrow U$  is a linear transformation with matrix

$$M_{g \circ f}(B_V, B_U) = M_g(B_W, B_U)M_f(B_V, B_W).$$

If  $V$  is finitely generated, a linear transformation  $f : V \rightarrow W$  is an isomorphism if and only if

$$\dim V = \dim(\text{Im}(f)) = \dim W.$$

Then, if  $\dim V = n$  the matrix  $M_f(B_V, B_W)$  is squared of size  $n \times n$  and

$$\text{rank}(M_f(B_V, B_W)) = \text{rank}(\text{Im}(f)) = \dim(\text{Im}(f)) = n,$$

$M_f(B_V, B_W)$  has nonzero determinant and therefore it is an invertible matrix.

The **inverse**  $f^{-1} : W \rightarrow V$  is also an isomorphism with matrix

$$M_{f^{-1}}(B_W, B_V) = (M_f(B_V, B_W))^{-1}.$$

**Example** Let us consider the linear transformations  $f, g : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ ,

$$f(x, y, z) = (4x - y, z + x, x), \quad g(x, y, z) = (y, 2z + 3x, z).$$

We obtain next the matrices of  $f - 2g$ ,  $f \circ g$ ,  $g \circ f$  with respect to the standard basis  $B$  of  $\mathbb{R}^3$ .

$$M_f(B) = \begin{pmatrix} 4 & -1 & 0 \\ 1 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix}, \quad M_g(B) = \begin{pmatrix} 0 & 1 & 0 \\ 3 & 0 & 2 \\ 0 & 0 & 1 \end{pmatrix},$$

$$M_{f-2g}(B) = M_f(B) - 2M_g(B) = \begin{pmatrix} 4 & -3 & 0 \\ -5 & 0 & -3 \\ 1 & 0 & -2 \end{pmatrix},$$

$$M_{f \circ g}(B) = M_f(B)M_g(B) = \begin{pmatrix} -3 & 4 & -2 \\ 0 & 1 & 1 \\ 0 & 1 & 0 \end{pmatrix},$$

$$M_{g \circ f}(B) = M_g(B)M_f(B) = \begin{pmatrix} 1 & 0 & 1 \\ 14 & -3 & 0 \\ 1 & 0 & 0 \end{pmatrix}.$$

This shows that the composition of linear transformations is not commutative in general.

Furthermore  $f$  and  $g$  are invertible linear transformations since

$$\det(M_f(B)) \neq 0 \text{ and } \det(M_g(B)) \neq 0.$$

The inverse linear transformations  $f^{-1}, g^{-1} : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  have matrices

$$M_{f^{-1}}(B) = \begin{pmatrix} 0 & 0 & 1 \\ -1 & 0 & 4 \\ 0 & 1 & -1 \end{pmatrix}, \quad M_{g^{-1}}(B) = \begin{pmatrix} 0 & 0 & -2/3 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$



# VECTOR INTERPRETATION OF A SYSTEM OF LINEAR EQUATIONS

Given a system of linear equations in the variables  $x_1, x_2, \dots, x_n$

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2 \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m \end{cases}$$

where  $m \geq 1$ ,  $a_{ij}, b_i \in \mathbb{R}$ ,  $i = 1, \dots, m$ ,  $j = 1, \dots, n$  and matrix equation  $AX = b$

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}, \quad X = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, \quad b = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix},$$

with  $A \in \mathcal{M}_{m \times n}(\mathbb{R})$  and  $b \in \mathcal{M}_{m \times 1}(\mathbb{R})$ .

A solution  $(s_1, \dots, s_n)$  of the system verifies

$$s_1 \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix} + s_2 \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix} + \dots + s_n \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix}.$$

That is  $(b_1, \dots, b_m)$  is a linear combination of the column vectors of the coefficient matrix of the system.

Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be the linear transformation with matrix  $A$  in the standard basis of  $\mathbb{R}^n$  and  $\mathbb{R}^m$ . The following items hold:

1.  $(b_1, \dots, b_m) \in \text{Im}(f) \Leftrightarrow AX = b$  has a solution.
2. Let  $S = (s_1, \dots, s_n) \in \mathbb{R}^n$  be a solution of  $AX = b$ . The solution set of  $AX = b$  equals the set of vectors whose image is  $(b_1, \dots, b_m)$

$$f^{-1}(b_1, \dots, b_m) = S + \text{Ker}(f),$$

where  $\text{Ker}(f)$  is the set of solutions of the system  $AX = 0$ .

**Example** Let  $f : \mathbb{R}^4 \rightarrow \mathbb{R}^2$  be the linear transformation whose matrix in the standard basis of  $\mathbb{R}^4$  and  $\mathbb{R}^2$  is

$$A = \begin{pmatrix} 2 & 0 & -1 & 2 \\ 1 & -3 & 2 & 1 \end{pmatrix}.$$

The vectors whose image is  $v = (1, -1) \in \mathbb{R}^2$  are the vectors in the set  $f^{-1}(v)$ , the solutions of

$$\begin{pmatrix} 2 & 0 & -1 & 2 \\ 1 & -3 & 2 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \end{pmatrix},$$

that is

$$(0, -1/3, -1, 0) + \langle (3, 5, 6, 0), (0, 5, 6, 3) \rangle$$

where  $(0, -1/3, -1, 0)$  is a vector of  $f^{-1}(v)$  and

$$\text{Ker}(f) = \langle (3, 5, 6, 0), (0, 5, 6, 3) \rangle.$$

## CHANGE OF COORDINATES

Assume that  $\dim V = n$ . Let  $B = \{v_1, \dots, v_n\}$  and  $B' = \{v'_1, \dots, v'_n\}$  be bases of  $V$ . Given a vector  $v \in V$ , denote by  $(x_1, \dots, x_n)_B$  its coordinates w.r.t.  $B$  and  $(x'_1, \dots, x'_n)_{B'}$  its coordinates w.r.t.  $B'$ .

Assume that the coordinates of the vectors of the basis  $B'$  w.r.t. the basis  $B$  are known, namely

$$\begin{aligned}v'_1 &= a_{11}v_1 + a_{21}v_2 + \cdots + a_{n1}v_n, \\v'_2 &= a_{12}v_1 + a_{22}v_2 + \cdots + a_{n2}v_n, \\&\vdots \\v'_n &= a_{1n}v_1 + a_{2n}v_2 + \cdots + a_{nn}v_n.\end{aligned}$$

Then, it holds

$$\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix} \begin{pmatrix} x'_1 \\ x'_2 \\ \vdots \\ x'_n \end{pmatrix}.$$

The matrix  $(a_{ij})$  is the **change of coordinates matrix** from  $B'$  to  $B$ : the matrix whose columns are the coordinates of the vectors in  $B'$  w.r.t.  $B$ . It is denoted by  $M(B', B)$ .

The change of coordinates matrix from  $B$  to  $B'$  is denoted by  $M(B, B')$  and it verifies

$$M(B, B') = M(B', B)^{-1}$$

then

$$\begin{pmatrix} x'_1 \\ x'_2 \\ \vdots \\ x'_n \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}^{-1} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}.$$

**Remark** The matrix  $M(B', B)$  is the matrix of an isomorphism from  $\mathbb{R}^n$  to  $\mathbb{R}^n$ .

**Example** Let  $B$  be the standard basis of  $\mathbb{R}^3$  and let

$$B' = \{(1, 3, 0), (1, 0, 2), (0, 4, -2)\}$$

be another basis of  $\mathbb{R}^3$ . Then, the matrix of the change of coordinates from  $B'$  to  $B$  is

$$M(B', B) = \begin{pmatrix} 1 & 1 & 0 \\ 3 & 0 & 4 \\ 0 & 2 & -2 \end{pmatrix}.$$

Let  $v$  be a vector with coordinates  $(1, 1, 2)_{B'}$ , its coordinates in the basis  $B$  are  $(2, 11, -2)_B$  and they are obtained by:

$$\begin{pmatrix} 2 \\ 11 \\ -2 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 0 \\ 3 & 0 & 4 \\ 0 & 2 & -2 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 2 \end{pmatrix}.$$



## EQUIVALENT MATRICES

Let us assume that  $\dim V = n$  and  $\dim W = m$ .

**Proposition** Let  $f : V \rightarrow W$  be a linear transformation. Given  $B_V$  and  $B'_V$  of  $V$ , and  $B_W$  and  $B'_W$  of  $W$ , it holds

$$M_f(B'_V, B'_W) = M(B_W, B'_W)M_f(B_V, B_W)M(B'_V, B_V).$$

**Definition** Two matrices  $A, A' \in \mathcal{M}_{m \times n}(\mathbb{R})$  are **equivalent** if there exist invertible matrices  $P \in \mathcal{M}_{n \times n}(\mathbb{R})$  and  $Q \in \mathcal{M}_{m \times m}(\mathbb{R})$  such that

$$A' = Q^{-1}AP.$$

Equivalently,  $A$  and  $A'$  are matrices of the same linear transformation  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$  in different basis.

**Example** Let  $f : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be the matrix of a linear transformation in the standard basis  $B_3$  of  $\mathbb{R}^3$  and  $B_2$  of  $\mathbb{R}^2$

$$M_f(B_3, B_2) = \begin{pmatrix} 3 & 0 & -2 \\ -1 & 4 & 5 \end{pmatrix}.$$

Also consider  $B'_3 = \{(1, 3, 0), (1, 0, 2), (0, 4, -2)\}$  of  $\mathbb{R}^3$  and  $B'_2 = \{(2, 1), (4, 3)\}$  of  $\mathbb{R}^2$ .

We have

$$M_f(B'_3, B'_2) = M(B_2, B'_2)M_f(B_3, B_2)M(B'_3, B_3),$$

This is

$$\begin{aligned} M_f(B'_3, B'_2) &= \begin{pmatrix} 2 & 4 \\ 1 & 3 \end{pmatrix}^{-1} \begin{pmatrix} 3 & 0 & -2 \\ -1 & 4 & 5 \end{pmatrix} \begin{pmatrix} 1 & 1 & 0 \\ 3 & 0 & 4 \\ 0 & 2 & -2 \end{pmatrix} \\ &= \begin{pmatrix} -35/2 & -39/2 & -6 \\ 19/2 & 19/2 & 4 \end{pmatrix}. \end{aligned}$$

**Proposition** Let  $f : V \rightarrow V$  be an endomorphism. Given bases  $B_V$  and  $B'_V$  of  $V$ . Then

$$M_f(B'_V) = M(B_V, B'_V)M_f(B_V)M(B'_V, B_V).$$

Two matrices  $A, A' \in \mathcal{M}_{n \times n}(\mathbb{R})$  verifying

$$A' = P^{-1}AP,$$

for some invertible matrix  $P \in \mathcal{M}_{n \times n}(\mathbb{R})$ , are matrices of an endomorphism  $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$  in two different bases.

**Example** Let  $f : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  be an endomorphism whose matrix in the standard basis  $B$  of  $\mathbb{R}^3$  is

$$M_f(B) = \begin{pmatrix} 2 & 0 & 1 \\ 0 & 1 & 1 \\ -1 & 1 & 3 \end{pmatrix}.$$

We also consider the basis  $B' = \{(1, 1, 0), (1, 0, 1), (0, 1, 1)\}$  of  $\mathbb{R}^3$ .

Then

$$M_f(B') = M(B', B)^{-1}M_f(B)M(B', B),$$

that is

$$\begin{aligned} M_f(B') &= \begin{pmatrix} 1/2 & 1/2 & -1/2 \\ 1/2 & -1/2 & 1/2 \\ -1/2 & 1/2 & 1/2 \end{pmatrix} \begin{pmatrix} 2 & 0 & 1 \\ 0 & 1 & 1 \\ -1 & 1 & 3 \end{pmatrix} \begin{pmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{pmatrix} \\ &= \begin{pmatrix} 3/2 & 1 & -1/2 \\ 1/2 & 2 & 3/2 \\ -1/2 & 0 & 5/2 \end{pmatrix}. \end{aligned}$$