

# Lesson 9

## Linear Map

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# Approach

**You cannot teach a man anything, you can only help him find it within himself.**







—Galileo Galilei

## Quotable Quote

**I never teach my pupils. I only attempt to provide the conditions in which they can learn.**

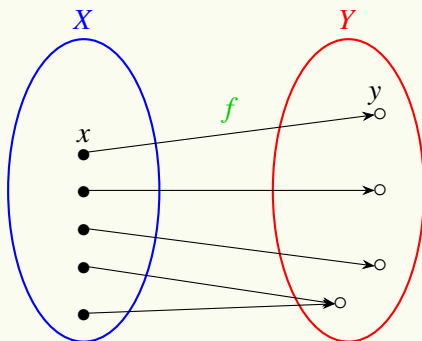
— Albert Einstein

# Learning Outcomes

-  Describe and compare different types of maps: onto (surjective), one-to-one (injective), bijective, composite, identity, inverse, and isomorphic maps.
-  Identify the characteristics of a linear map and examine its relationship with vectors and matrices.
-  Evaluate whether a transformation is linear or prove that a transformation is linear.
-  Interpret the concept of representation matrix of a map.
-  Describe the relationships between image, kernel, dimension of the linear space, the rank of a matrix, and how the rank is connected to injection and surjection.
-  Appraise and connect the concept of isomorphic map to basis transformation and the regularity of a transformation matrix.

# Map

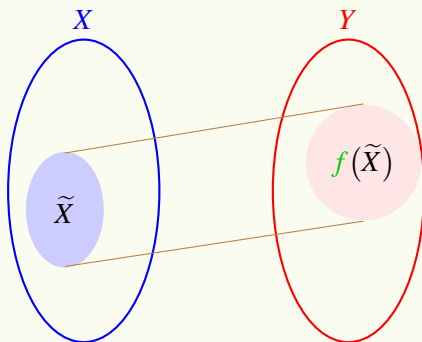
📍 Mapping or **map**  $f : X \rightarrow Y$ .



📍 Example: ISOweekday  $X = \{\text{Sunday, Monday, } \dots, \text{Saturday}\}$ , and  $Y = \{0, 1, \dots, 6\}$ .

## Mapping of Subset and Image

🔑 The **image**  $f(\tilde{X}) = \{f(x) \mid x \in \tilde{X}\}$ .



🔑 Example:  $\tilde{X} = [0, 2\pi] \subset \mathfrak{R} = X$  and  $y = f(x) = \sin(x)$ . Hence,  $f(\tilde{X}) = [-1, 1] \subset \mathfrak{R} = Y$ .

## Domain and Range (Image)

- 🔗 Let  $T : U \rightarrow V$  be a map.
- 🔗  $U$  is called the **domain** of  $T$  and  $V$  the **codomain**.
- 🔗 The **range** or **image** of  $T$ , denoted by  $\text{range}(T)$ , is the set of all possible outputs in the codomain.
- 🔗 Mathematically,  $\text{range}(T) = \text{image}(T) = \{T(x) : x \in U\}$ .
- 🔗 For example, if  $T$  is given by  $T(\mathbf{x}) = \mathbf{A}\mathbf{x}$  for some matrix  $\mathbf{A}$ , then the range of  $T$  is given by the **span (set of all possible linear combinations) of  $\mathbf{A}$ 's column vectors**.

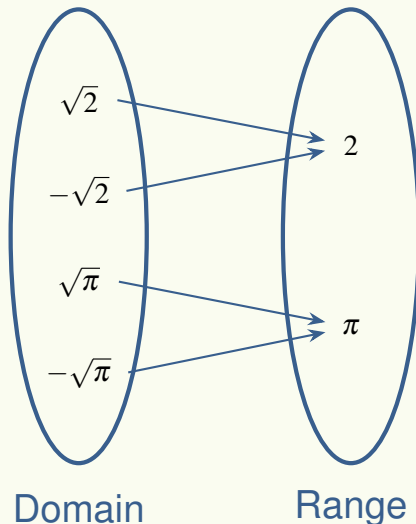
# Surjection

- 🔗  $T$  is said to be **surjective** (or **onto**) if its range equals the codomain.
- 🔗 It means that every vector in the codomain can be the output of  $T$ .
- 🔗 If  $T$  is surjective, it is called a **surjection**.

## 🔗 Example

- Let  $T \left( \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \right) = \begin{bmatrix} x_1 - x_2 \\ -x_1 + x_2 \end{bmatrix}$ .
- Note that every vector in  $\text{range}(T)$  is of the form  $\begin{bmatrix} a \\ -a \end{bmatrix}$ .
- Hence,  $T$  is not surjective since there are vectors in  $\mathfrak{R}^2$  that are not in the range of  $T$ , e.g.  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ .

# Illustration of Surjection



## Injection and Bijection

🔗  $T$  is said to be **injective** (or **one-to-one**) if for all distinct  $\mathbf{x}, \mathbf{y} \in U$ ,  $T(\mathbf{x}) \neq T(\mathbf{y})$ .

🔗 It means that different inputs lead to different outputs.

🔗 If  $T$  is injective, it is called an **injection**.

🔗 Example

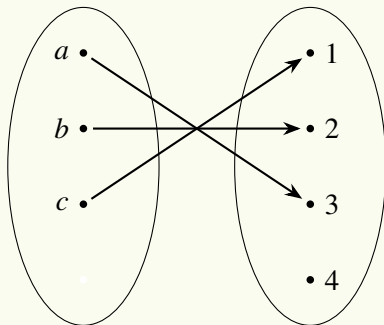
- Consider the same  $T$  in the example above. It is not injective because for every  $a \in \mathfrak{R}$ ,

$$T\left(\begin{bmatrix} a \\ a \end{bmatrix}\right) = \begin{bmatrix} a - a \\ -a + a \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

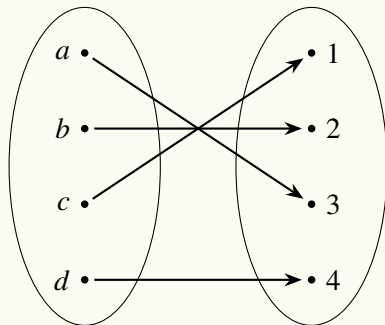
🔗 If  $T$  is both surjective and injective, it is said to be **bijective** and we call  $T$  a **bijection**.

# Illustration of Injection and Bijection

## Injection



## Bijection



## Some Important Maps

📌 The **identity map**:  $f(x) = x$  for all  $x \in X$ .

📌 A **composite map**: Let  $f : X \rightarrow Y$  and  $g : Y \rightarrow Z$ . Then  $g \circ f : X \rightarrow Z$  is a composite map. In other words,

$$(g \circ f)(x) = g(f(x)).$$

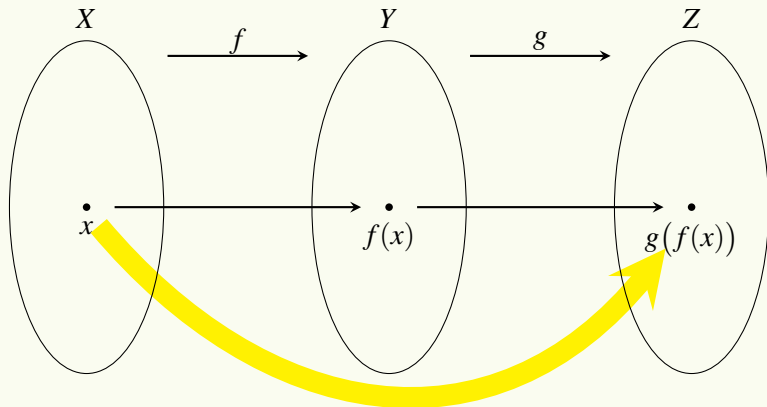
📌 The **inverse map**: Let  $f : X \rightarrow Y$ . Then for  $y \in Y$ , the set of  $x \in X$  that satisfies  $f(x) = y$  is called the inverse image, and it is represented by  $f^{-1}(y)$ . That is

$$f^{-1}(y) = \left\{ x \in X \mid f(x) = y \right\}.$$

📌 If  $f : X \rightarrow Y$  is bijective, there is only one inverse image that corresponds to every  $y \in Y$ . Then we have the inverse map from  $Y$  back unto  $X$ , and it is expressed as  $f^{-1}$ , so that  $x = f^{-1}(y)$ . Moreover,  $f^{-1} : Y \rightarrow X$  is bijective too.

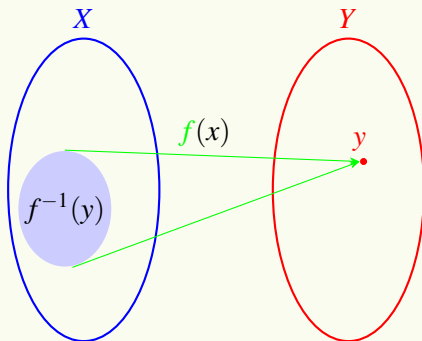
# Illustration of Composite Map

🔗 A **composite map**:  $(g \circ f)(x) = g(f(x))$ .



# Illustration of Inverse Map

🔗 The **inverse map**  $f^{-1}(y) = \{x \in X \mid f(x) = y\}$ .



## Definition of Linear Map

- Let  $V$  and  $W$  be vector spaces.
- A map  $T : V \rightarrow W$  is a **linear transformation** if for any two vectors  $\mathbf{x}$  and  $\mathbf{y}$  in  $V$  and any scalar  $c \in \mathfrak{R}$ , the following are satisfied:

$$T(\mathbf{x} + \mathbf{y}) = T(\mathbf{x}) + T(\mathbf{y}) \quad (1)$$

$$T(c\mathbf{x}) = cT(\mathbf{x}) \quad (2)$$

## Matrix as Linear Map

### Theorem 3.1.

Let  $\mathbf{A}$  be a  $m \times n$  matrix. Consider a map  $g$  from  $\mathfrak{R}^n$  to  $\mathfrak{R}^m$  determined by the matrix  $\mathbf{A}$ :

$$g : \mathbf{x} \longrightarrow \mathbf{y} = \mathbf{Ax} \quad (\mathbf{x} \in \mathfrak{R}^n; \mathbf{y} \in \mathfrak{R}^m).$$

The map  $g$  is a linear map.

### Proof.

🐟 For  $\mathbf{x}_1, \mathbf{x}_2 \in \mathfrak{R}^n$ , we have  $\mathbf{y}_1 = g(\mathbf{x}_1)$  and  $\mathbf{y}_2 = g(\mathbf{x}_2)$ .

🐟 Now,  $g(\mathbf{x}_1 + \mathbf{x}_2) = \mathbf{A}(\mathbf{x}_1 + \mathbf{x}_2) = \mathbf{Ax}_1 + \mathbf{Ax}_2 = \mathbf{y}_1 + \mathbf{y}_2$ .

🐟 Hence, (1) is satisfied.

🐟 Next, let  $c$  be the scalar. (2) is satisfied as well:

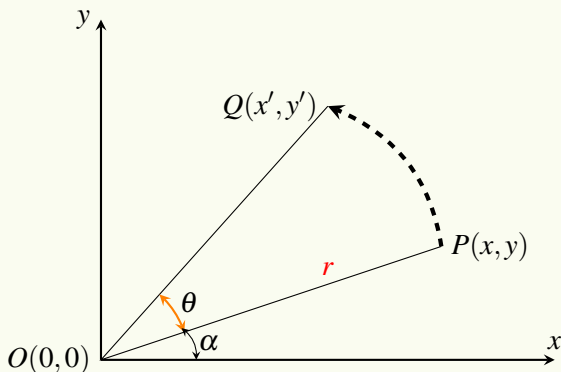
$$g(c\mathbf{x}) = \mathbf{A}(c\mathbf{x}) = c(\mathbf{Ax}) = cg(\mathbf{x}).$$



## Example: Rotation Matrix

### Example 3.2.

➤ A point  $P(x,y)$  on a plane is rotated  $\theta$  degrees anticlockwise with respect to the origin  $O(0,0)$ . Is  $\begin{bmatrix} x' \\ y' \end{bmatrix} = f\left(\begin{bmatrix} x \\ y \end{bmatrix}\right)$  a linear map?



## Answer to Example 3.2

🐟 In the polar coordinate system,  $x$  and  $y$  can be expressed in terms of the radius  $r$  and the angle  $\alpha$ :

$$x = r \cos(\alpha), \quad y = r \sin(\alpha).$$

🐟 It follows that

$$\begin{aligned} \begin{bmatrix} x' \\ y' \end{bmatrix} &= \begin{bmatrix} r \cos(\alpha + \theta) \\ r \sin(\alpha + \theta) \end{bmatrix} = \begin{bmatrix} r \cos(\alpha) \cos(\theta) - r \sin(\alpha) \sin(\theta) \\ r \sin(\theta) \cos(\alpha) + r \cos(\theta) \sin(\alpha) \end{bmatrix} \\ &= \begin{bmatrix} \cos(\theta)x - \sin(\theta)y \\ \sin(\theta)x + \cos(\theta)y \end{bmatrix} \\ &= \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}. \end{aligned}$$

🐟 According to Theorem 3.1,  $f: \mathfrak{R}^2 \rightarrow \mathfrak{R}^2$  is a linear map through the **rotation matrix**  $\mathbf{R}(\theta) := \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$ .

## Example: Projected Rotation

### Example 3.3.

Let  $(a, b, c)$  be the coordinates of a point  $P$  in  $\mathfrak{R}^3$ .

Then, we define  $\mathbf{u}_1 = \begin{bmatrix} \cos(\alpha) \\ \sin(\alpha) \end{bmatrix}$ ,  $\mathbf{u}_2 = \begin{bmatrix} \cos(\beta) \\ \sin(\beta) \end{bmatrix}$ , and  $\mathbf{u}_3 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ .

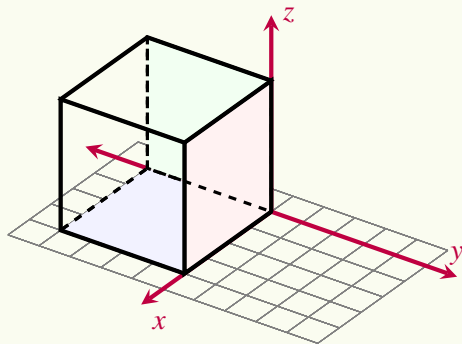
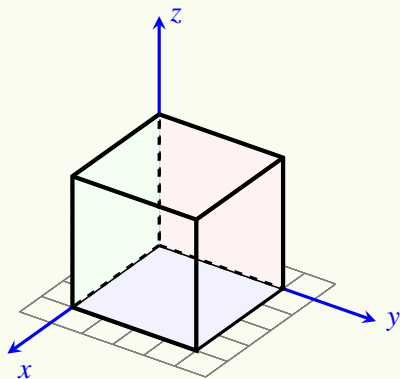
Let us define a point  $Q$  in  $\mathfrak{R}^2$  by

$$\overrightarrow{OQ} = a\mathbf{u}_1 + b\mathbf{u}_2 + c\mathbf{u}_3.$$

Let  $(p, q)$  be the coordinates of point  $Q$ .

Is the map from  $(a, b, c)$  to  $(p, q)$  linear?

# Illustration of Projected Rotation



## Answer to Example 3.3

- 🐟 In the matrix form, the relationship between  $(a, b, c)$  and  $(p, q)$  is given by

$$\begin{bmatrix} p \\ q \end{bmatrix} = [\mathbf{u}_1 \quad \mathbf{u}_2 \quad \mathbf{u}_3] \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} \cos(\alpha) & \cos(\beta) & 0 \\ \sin(\alpha) & \sin(\beta) & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

- 🐟 Define  $\mathbf{A} := \begin{bmatrix} \cos(\alpha) & \cos(\beta) & 0 \\ \sin(\alpha) & \sin(\beta) & 1 \end{bmatrix}$ .

- 🐟 It follows that

$$\begin{bmatrix} p \\ q \end{bmatrix} = \mathbf{A} \begin{bmatrix} a \\ b \\ c \end{bmatrix}.$$

- 🐟 Applying Theorem 3.1, we conclude that the map  $f : \mathfrak{R}^3 \rightarrow \mathfrak{R}^2$  is a linear map.

# Isomorphic Map





## Definition 3.4.

Let  $f$  be a linear map from the linear space  $U$  to another linear space  $V$ . Moreover  $f$  is assumed to be bijective. When these conditions are fulfilled,  $f$  is said to be an **isomorphic map**, and  $U$  and  $V$  are said to be **isomorphic**.


## Theorem 3.5.

*When the linear spaces  $U$  and  $V$  have the same dimension, then there exists an **isomorphic map** that connects these two spaces.*

## Proof of Theorem 3.5

-  Let  $\dim U = \dim V = n$ , and let their respective basis be  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$  and  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ .
-   $U$  being a linear space, we can write  $\mathbf{x} = c_1\mathbf{u}_1 + \dots + c_n\mathbf{u}_n$ .
-  Let  $f : U \rightarrow V$  be a linear map, so that  $f(\mathbf{u}_i) = \mathbf{v}_i$ , where  $i = 1, \dots, n$ . Consequently,  $f(\mathbf{x}) = c_1\mathbf{v}_1 + \dots + c_n\mathbf{v}_n$ .
-  Suppose  $U \ni \mathbf{y} = d_1\mathbf{u}_1 + \dots + d_n\mathbf{u}_n$ . Then

$$\begin{aligned}
 f(\mathbf{x} + \mathbf{y}) &= (c_1 + d_1)\mathbf{v}_1 + \dots + (c_n + d_n)\mathbf{v}_n \\
 &= (c_1\mathbf{v}_1 + \dots + c_n\mathbf{v}_n) + (d_1\mathbf{v}_1 + \dots + d_n\mathbf{v}_n) \\
 &= f(\mathbf{x}) + f(\mathbf{y}).
 \end{aligned}$$

-  Also, for  $a \in \mathfrak{R}$ ,
 
$$f(a\mathbf{x}) = ac_1\mathbf{v}_1 + \dots + ac_n\mathbf{v}_n = a(c_1\mathbf{v}_1 + \dots + c_n\mathbf{v}_n) = af(\mathbf{x}).$$

## Proof of Theorem 3.5 (Cont'd)

- For all  $\mathbf{x} \in U$ , the range  $\{f(\mathbf{x}) | \mathbf{x} \in U\}$  is  $V$ . So  $f$  is surjective.
- Moreover, we let  $f(\mathbf{x}) = f(\mathbf{y})$ . By definition,  $c_1 = d_1, \dots, c_n = d_n$ , and it follows that  $\mathbf{x} = \mathbf{y}$ , which means that  $f$  is injective.
- Being both surjective and injective,  $f$  is therefore bijective.
- The conditions in Definition 3.4 are satisfied, which means that  $f$  is isomorphic.



## Example of an Isomorphic Map

### Example 3.6.

Consider the set of quadratic functions:

$$\mathbb{R}[x]_2 = \{a_0 + a_1x + a_2x^2 \mid a_0, a_1, a_2 \in \mathfrak{R}\}.$$

The standard basis of  $\mathbb{R}[x]_2$  is  $\{1, x, x^2\}$ .

It is clear that  $\dim \mathbb{R}[x]_2 = \dim \mathfrak{R}^3 = 3$ .

By Theorem 3.5, there exists an isomorphic map that connects  $\mathbb{R}[x]_2$  and  $\mathfrak{R}^3$ .

Concretely,  $p = s + tx + ux^2 \in \mathbb{R}[x]_2$  and

$$f(p) = se_1 + te_2 + ue_3 = \begin{bmatrix} s \\ t \\ u \end{bmatrix} \in \mathfrak{R}^3.$$

## Representation Matrix of a Map

### Definition 3.7 (Representation Matrix of a Map).

- Let  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$  and  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m\}$  be the respective basis of linear spaces  $U$  and  $V$ , and  $f : U \rightarrow V$  is a linear map.
- When  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$  are mapped by  $f$ , they become the elements of  $V$  and thus can be written as the linear combinations of  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$ . That is,

$$f(\mathbf{u}_1) = a_{11}\mathbf{v}_1 + a_{21}\mathbf{v}_2 + \cdots + a_{m1}\mathbf{v}_m,$$

$$f(\mathbf{u}_2) = a_{12}\mathbf{v}_1 + a_{22}\mathbf{v}_2 + \cdots + a_{m2}\mathbf{v}_m,$$

$$\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$$

$$f(\mathbf{u}_n) = a_{1n}\mathbf{v}_1 + a_{2n}\mathbf{v}_2 + \cdots + a_{mn}\mathbf{v}_m.$$

- Let  $\mathbf{A} = [a_{ij}]_{m \times n}$ . It is called the **representation matrix** with respect to  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$  and  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m\}$ , and the **matrix corresponding to the linear map  $f$** .

## Example

### Example 3.8.

Find the matrix  $\mathbf{A}$  that corresponds to the map  $f$  in Example 3.2.

➤ The standard basis of  $\mathfrak{R}^2$  is  $\mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $\mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ .

➤ Accordingly,

$$f(\mathbf{e}_1) = \cos(\theta)\mathbf{e}_1 + \sin(\theta)\mathbf{e}_2, \quad f(\mathbf{e}_2) = -\sin(\theta)\mathbf{e}_1 + \cos(\theta)\mathbf{e}_2.$$

➤ It follows that

$$[f(\mathbf{e}_1) \quad f(\mathbf{e}_2)] = [\mathbf{e}_1 \quad \mathbf{e}_2] \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}.$$

➤ Thus, we have found  $\mathbf{A}$  to be the rotation matrix

$$\mathbf{R}(\theta) = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} \text{ in Example 3.2.}$$

# What is image? What is kernel?

Let  $f$  be a linear map from the linear space  $U$  to another linear space  $V$ .

└ The **image** of  $f$  is the set

$$\text{Im } f := \{f(\mathbf{u}) \mid \mathbf{u} \in U\} \quad (= f(U)),$$

which is a subspace of  $V$ .

└ The **kernel** is the set

$$\text{Ker } f := \{\mathbf{u} \in U \mid f(\mathbf{u}) = \mathbf{0}\},$$

which is a subspace of  $U$ .

# Examples of Image and Kernel

## Image

- ✍ In Example 3.2, all the points of  $\mathfrak{R}^2$  can be rotated. Therefore  $\text{Im}f = \{(x, y) | x, y \in \mathfrak{R}\} = \mathfrak{R}^2$ .
  - ✍ All the points of  $\mathfrak{R}^2$  can be mapped to the  $x$  axis by a linear map  $f$ . Therefore,  $\text{Im}f = \{(x, 0) | x \in \mathfrak{R}\}$ .
- 

## Kernel

- ✍ In Example 3.2, only the origin can be mapped to the origin. Therefore  $\text{Ker}f = \{(0, 0)\}$ .
- ✍ All the points of the  $y$  axis are mapped to the origin but not the other points in  $\mathfrak{R}^2$ . Therefore,  $\text{Ker}f = \{(0, y) | f(0, y) = (0, 0), \forall y \in \mathfrak{R}\}$ .

## Theorems

✧  $\text{Ker} f = \{\mathbf{0}\} \iff f$  is bijective.

✧  $\dim \text{Ker} f + \dim \text{Im} f = \dim U$

✧ With respect to  $m \times n$  matrix  $\mathbf{A}$  and  $n \times \ell$  matrix  $\mathbf{B}$ ,

$$\text{rank} \mathbf{AB} \leq \text{rank} \mathbf{A}, \quad \text{and} \quad \text{rank} \mathbf{AB} \leq \text{rank} \mathbf{B}.$$

✧ With respect to  $m \times n$  matrix  $\mathbf{A}$ ,  $m$ -dimensional regular square matrix  $\mathbf{P}$ , and  $n$ -dimensional regular square matrix  $\mathbf{Q}$ ,

$$\text{rank} \mathbf{PA} = \text{rank} \mathbf{A} = \text{rank} \mathbf{AQ}.$$

✧ Let  $\mathbf{A}$  be the matrix that corresponds to the linear map  $f: \mathfrak{R}^n \rightarrow \mathfrak{R}^m$ . Then

①  $\text{rank} \mathbf{A} = n \iff f$  is injective.

②  $\text{rank} \mathbf{A} = m \iff f$  is surjective.

# Theorem on Basis Transformation

## Theorem 5.1.

- Let  $\mathbf{P}$  be the basis transformation matrix for the  $n$ -dimensional linear space  $U$ . It transforms the basis  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$  to another basis  $\{\tilde{\mathbf{u}}_1, \tilde{\mathbf{u}}_2, \dots, \tilde{\mathbf{u}}_n\}$ .
- Likewise,  $\mathbf{Q}$  denotes the basis transformation matrix for the  $m$ -dimensional linear space  $V$ . It transforms the basis  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m\}$  to another basis  $\{\tilde{\mathbf{v}}_1, \tilde{\mathbf{v}}_2, \dots, \tilde{\mathbf{v}}_m\}$ .
- $f$  is a linear map  $f : U \rightarrow V$  with  $\mathbf{A}$  being the corresponding representation matrix that transforms the basis of  $U$ ,  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ , to the basis of  $V$ ,  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m\}$ .
- Then, the matrix  $\tilde{\mathbf{A}}$  that corresponds to the transformation of  $\{\tilde{\mathbf{u}}_1, \tilde{\mathbf{u}}_2, \dots, \tilde{\mathbf{u}}_n\}$  to  $\{\tilde{\mathbf{v}}_1, \tilde{\mathbf{v}}_2, \dots, \tilde{\mathbf{v}}_m\}$  is

$$\tilde{\mathbf{A}} = \mathbf{Q}^{-1}\mathbf{A}\mathbf{P}.$$

# Isomorphic Map and Regularity of Matrix

## Theorem 5.2.

Let  $\mathbf{P}$  be the basis transformation matrix for the  $n$ -dimensional linear space  $U$ . It transforms the basis  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$  to another basis  $\{\tilde{\mathbf{u}}_1, \tilde{\mathbf{u}}_2, \dots, \tilde{\mathbf{u}}_n\}$ . Let  $f : U \rightarrow U$  be a map with  $\mathbf{A}$  being the corresponding matrix with respect to the basis  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ . The matrix that corresponds to the map  $f$  with respect to the basis  $\{\tilde{\mathbf{u}}_1, \tilde{\mathbf{u}}_2, \dots, \tilde{\mathbf{u}}_n\}$  is  $\tilde{\mathbf{A}}$ . Then

$$\tilde{\mathbf{A}} = \mathbf{P}^{-1}\mathbf{A}\mathbf{P}.$$

## Theorem 5.3.

Linear map  $f$  being isomorphic is equivalent to the matrix  $\mathbf{A}$  that corresponds to  $f$  being regular.

## Takeaways

- ✂ A map is a generalized version of function.
- ✂ A linear map from a space to another space provides a geometric meanings such as rotation and projection to its corresponding matrix that is closed under addition and scalar multiplication, i.e., linear combination.
- ✂ The inverse map is unique only for the bijective map.
- ✂ Isomorphism refers to two linear spaces having a bijection connecting every point of one space to the corresponding point in the other space.
- ✂ The kernel is in the input space, and the image is in the output space.
- ✂ Isomorphic map is equivalent to a regular matrix that corresponds to it.

# Keywords

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**bijjective, 11**

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