

Week 5 Linear Map

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Table of Contents

- 1 Introduction
- 2 Map
- 3 Linear Map
- 4 Image and Kernel
- 5 Different Basis
- 6 Takeaways

Approach

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

—Galileo Galilei

Quotable Quote

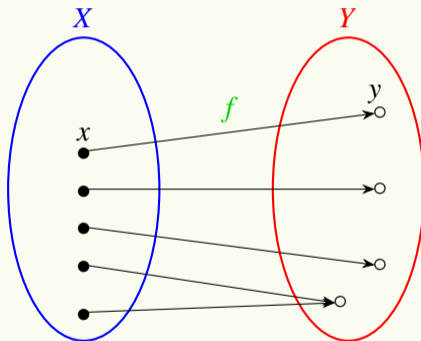
I never teach my pupils. I only attempt to provide the conditions in which they can learn.

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

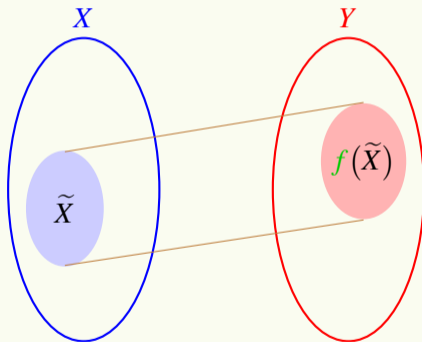
📍 Mapping or **map** $f : X \longrightarrow Y$.



📍 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, $f(\tilde{X}) = [-1, 1] \subset \mathfrak{R} = Y$.

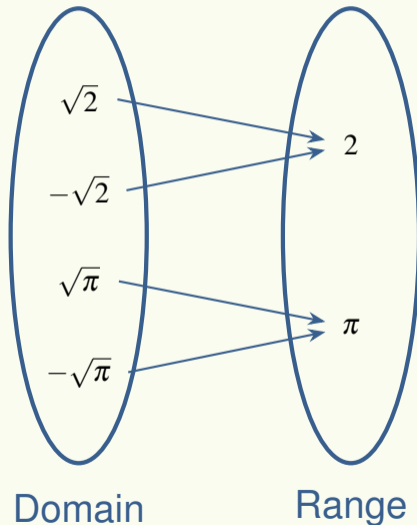
Domain and Range (Image)

- Let $T : U \longrightarrow V$ 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 is 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 $x, y \in U$, $T(x) \neq T(y)$.

🔗 It means that different inputs lead to different outputs.

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

🔗 Example

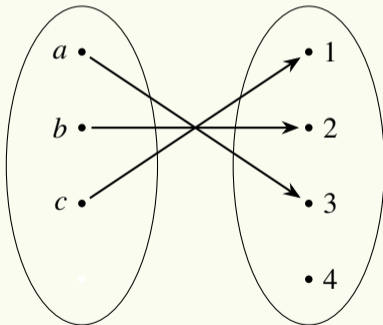
- Consider the same T in the example above. It is not injective because for every $a \in \mathfrak{R}$,

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

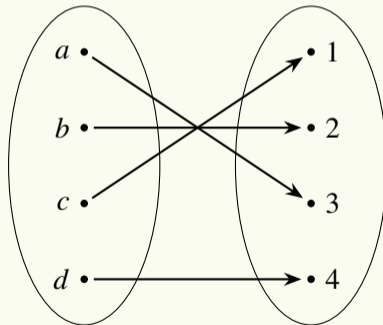
🔗 If T is both surjective and injective, it is said to be **bijective** and we call T a **bijection**.

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

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

🔗 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**: $(g \circ f)(x) = g(f(x))$.

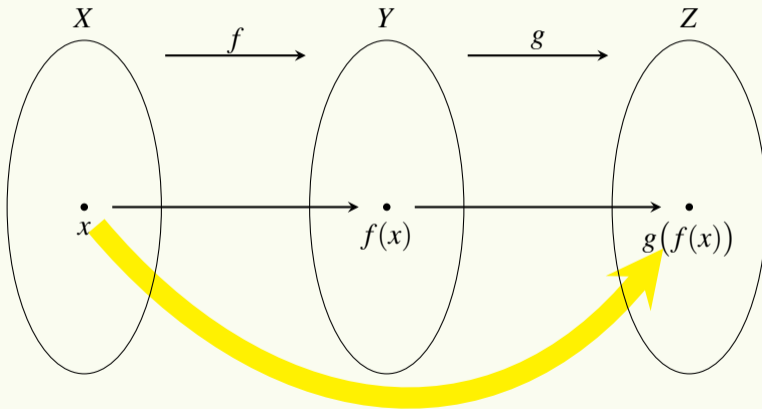
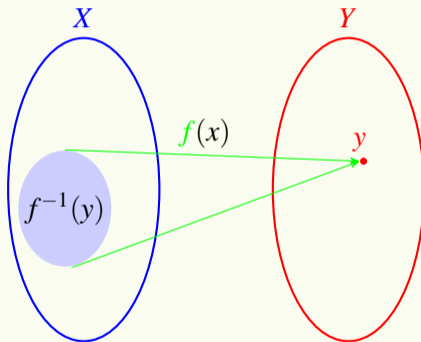


Illustration of Inverse Map

📍 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 $T : V \longrightarrow W$ 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 two conditions 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:

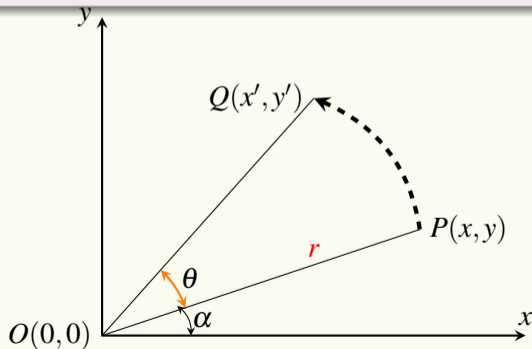
$$g(c\mathbf{x}) = \mathbf{A}(c\mathbf{x}) = c(\mathbf{Ax}) = cg(\mathbf{x}).$$



Example: Rotation Matrix

Example 3.2.

🐟 A point $P(x, y)$ on a plane is rotated θ 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 α :

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

🐟 It follows that

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

🐟 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 $\mathbf{u}_3 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$.

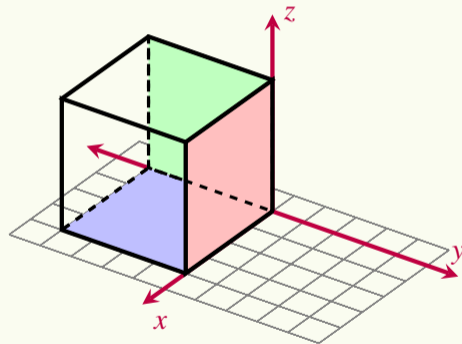
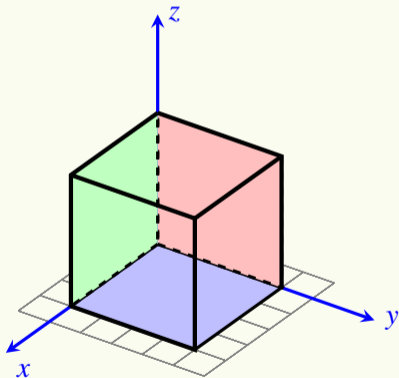
Let us define a point Q in \mathfrak{R}^2 by

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

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 $f : \mathfrak{R}^3 \longrightarrow \mathfrak{R}^2$ 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,
- $$\begin{aligned} f(\mathbf{u}_1) &= a_{11}\mathbf{v}_1 + a_{21}\mathbf{v}_2 + \cdots + a_{m1}\mathbf{v}_m, \\ f(\mathbf{u}_2) &= a_{12}\mathbf{v}_1 + a_{22}\mathbf{v}_2 + \cdots + a_{m2}\mathbf{v}_m, \\ &\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \\ f(\mathbf{u}_n) &= a_{1n}\mathbf{v}_1 + a_{2n}\mathbf{v}_2 + \cdots + a_{mn}\mathbf{v}_m. \end{aligned}$$
- 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 \mathcal{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 $\mathbf{R}(\theta) = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$ 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, $\text{Ker } f = \{(0, y) | f(0, y) = (0, 0), \forall y \in \mathfrak{R}\}$.

Example

└ We define linear mapping $f : \mathfrak{R}^q \longrightarrow \mathfrak{R}^p$ as

$$\mathbf{x} \longrightarrow \mathbf{y} = f(\mathbf{x}) = \mathbf{A}\mathbf{x},$$

where \mathbf{A} is $p \times q$ matrix, which is written as $[\mathbf{a}_1 \ \mathbf{a}_2 \ \cdots \ \mathbf{a}_q]$.

└ Then

$$\text{Im}f = f(\mathfrak{R}^q) = \{x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \cdots + x_n\mathbf{a}_n \mid x_1, x_2, \dots, x_n \in \mathfrak{R}\} = \langle \mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n \rangle$$

└ So,

$$\dim \text{Im}f = \text{rank } \mathbf{A}.$$

└ Now, $\text{Ker}f$ is identical to the solution set of $\mathbf{A}\mathbf{x} = \mathbf{0}$. Hence

$$\dim \text{Ker}f = n - \text{rank } \mathbf{A}.$$

Theorems

- ✎ $\text{Ker } f = \{\mathbf{0}\} \iff f$ is bijective.
- ✎ $\dim \text{Ker } f + \dim \text{Im } f = \dim V$ for linear map $f : V \rightarrow V'$.
- ✎ 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.

Example

- Define the linear map $f : \mathfrak{R}^3 \rightarrow \mathfrak{R}^2$ as

$$\mathbf{x} \rightarrow \mathbf{Ax}, \quad \mathbf{A} = \begin{bmatrix} 1 & -2 & 0 \\ -1 & 3 & -1 \end{bmatrix}.$$

- Find $\text{Ker}f$ and its dimension.

- Solution: First we simplify \mathbf{A} to $\mathbf{B} = \begin{bmatrix} 1 & 0 & -2 \\ 0 & 1 & -1 \end{bmatrix}$. With $\mathbf{x}' = [x_1 \ x_2 \ x_3]$ and $\mathbf{Ax} = \mathbf{0}$, we

obtain $\mathbf{x} = \begin{bmatrix} 2c \\ c \\ c \end{bmatrix}$.

- Therefore

$$\text{Ker}f = \left\{ c \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix} \mid c \in \mathfrak{R} \right\} \quad \dim \text{Ker}f = 1.$$

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

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

Proof of Theorem 5.1 (1 of 2)

By the definition of basis transformation matrices P and Q ,

$$(\tilde{\mathbf{u}}_1, \dots, \tilde{\mathbf{u}}_n) = (\mathbf{u}_1, \dots, \mathbf{u}_n)P, \quad (\tilde{\mathbf{v}}_1, \dots, \tilde{\mathbf{v}}_m) = (\mathbf{v}_1, \dots, \mathbf{v}_m)Q,$$

where P and Q are regular matrices.

Again by the definitions of A and \tilde{A} ,

$$(f(\mathbf{u}_1), \dots, f(\mathbf{u}_n)) = (\mathbf{v}_1, \dots, \mathbf{v}_m)A, \quad (f(\tilde{\mathbf{u}}_1), \dots, f(\tilde{\mathbf{u}}_n)) = (\tilde{\mathbf{v}}_1, \dots, \tilde{\mathbf{v}}_m)\tilde{A}.$$

By these definitions, we have

$$(f(\tilde{\mathbf{u}}_1), \dots, f(\tilde{\mathbf{u}}_n)) = (\tilde{\mathbf{v}}_1, \dots, \tilde{\mathbf{v}}_m)\tilde{A} = (\mathbf{v}_1, \dots, \mathbf{v}_m)Q\tilde{A}.$$

Proof of Theorem 5.1 (2 of 2)

📌 We express \mathbf{P} as $[\mathbf{p}_1 \ \cdots \ \mathbf{p}_n]$ and obtain

$$(\mathbf{u}_1, \dots, \mathbf{u}_n)\mathbf{P} = \left((\mathbf{u}_1, \dots, \mathbf{u}_n)\mathbf{p}_1, \dots, (\mathbf{u}_1, \dots, \mathbf{u}_n)\mathbf{p}_n \right)$$

📌 Knowing that f is a linear map, we have, for $j = 1, 2, \dots, n$,

$$f(\tilde{\mathbf{u}}_j) = f\left((\mathbf{u}_1, \dots, \mathbf{u}_n)\mathbf{p}_j \right) = (f(\mathbf{u}_1), \dots, f(\mathbf{u}_n))\mathbf{p}_j.$$

📌 It follows that $(f(\tilde{\mathbf{u}}_1), \dots, f(\tilde{\mathbf{u}}_n)) = (f(\mathbf{u}_1), \dots, f(\mathbf{u}_n))\mathbf{P} = (\mathbf{v}_1, \dots, \mathbf{v}_m)\mathbf{A}\mathbf{P}$.

📌 Finally, since $\mathbf{v}_1, \dots, \mathbf{v}_m$ are linearly independent, we must have

$$\mathbf{Q}\tilde{\mathbf{A}} = \mathbf{A}\mathbf{P}.$$



Isomorphic Map and Regularity of Matrix

Theorem 5.2.

Let 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 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{A} . Then

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

Theorem 5.3.

Linear map f being isomorphic is equivalent to the matrix A that corresponds to f being regular.

Example

Let the standard basis transformation matrix be $\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 4 & -1 \end{bmatrix}$. The other basis is defined by $\mathbf{u}'_1 = \begin{bmatrix} 1 & 1 \end{bmatrix}$ and $\mathbf{u}'_2 = \begin{bmatrix} -1 & 2 \end{bmatrix}$.

Find the basis transformation matrix \mathbf{P} that transforms the standard basis to the basis $\{\mathbf{u}_1, \mathbf{u}_2\}$, and also its correspondence matrix $\tilde{\mathbf{A}}$.

Solution:

$$[\mathbf{u}_1 \ \mathbf{u}_2] = [\mathbf{e}_1 \ \mathbf{e}_2]\mathbf{P},$$

which is

$$\begin{bmatrix} 1 & -1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}\mathbf{P}.$$

Since $\mathbf{P}^{-1} = \frac{1}{3} \begin{bmatrix} 2 & 1 \\ -1 & 1 \end{bmatrix}$, and by Theorem 5.2, we obtain

$$\tilde{\mathbf{A}} = \mathbf{P}^{-1}\mathbf{A}\mathbf{P} = \begin{bmatrix} 3 & 0 \\ 0 & -3 \end{bmatrix}.$$

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

bijection, 11

bijective, 11

codomain, 8

composite map, 13, 14

domain, 8

identity map, 13

image, 7, 8, 29

injection, 11

injective, 11

inverse map, 13, 15

isomorphic, 23

isomorphic map, 23

kernel, 29

linear transformation, 16

map, 6

**matrix corresponding to the linear map
 f , 27**

one-to-one, 11

onto, 9

range, 8

representation matrix, 27

rotation matrix, 19

surjection, 9

surjective, 9