

Linear Maps

1 Linear Maps

1.1 Definition

Let V and W be vector spaces. A function $f : V \rightarrow W$ is called a linear map, or a linear transformation from V to W if, for all $u, v \in V, k \in \mathbb{R}$

$$f(\mathbf{u} + \mathbf{v}) = f(\mathbf{u}) + f(\mathbf{v}) \text{ and } f(k\mathbf{u}) = k \cdot f(\mathbf{u})$$

If $V=W$ then f is called a linear operator

1.2 Examples

- The map $f : V \rightarrow W$ such that $f(u) = 0$ for all u is linear
- If A is an $m \times n$ matrix then the map $f_A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ given by $f_A(x) = Ax$ is linear. (Here x and Ax are column vectors in \mathbb{R}^n and \mathbb{R}^m , respectively)
Indeed,

$$f_A(\mathbf{u} + \mathbf{v}) = A(\mathbf{u} + \mathbf{v}) = A\mathbf{u} + A\mathbf{v} = f_A(\mathbf{u}) + f_A(\mathbf{v})$$

and

$$f_A(k\mathbf{u}) = A(k\mathbf{u}) = k(A\mathbf{u}) = kf_A(\mathbf{u})$$

2 Non-examples in \mathbb{R}^2

- The map $g : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ defined by $g(x, y) = (x, y + 1)$ is not linear. Indeed, any linear map f satisfies

$$f(\mathbf{0}) = f(0\mathbf{x}) = 0f(\mathbf{x}) = \mathbf{0}$$

and the above map g fails this property

- The map $f : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ defined by $f(x, y) = (0, xy)$ is not linear. Indeed,

$$f(e_1 + e_2) = f(1, 1) = (0, 1)$$

while

$$f(e_1) + f(e_2) = f(0, 1) + f(1, 0) = (0, 0)$$

3 Example in \mathbb{R}^2

3.1 Reflection

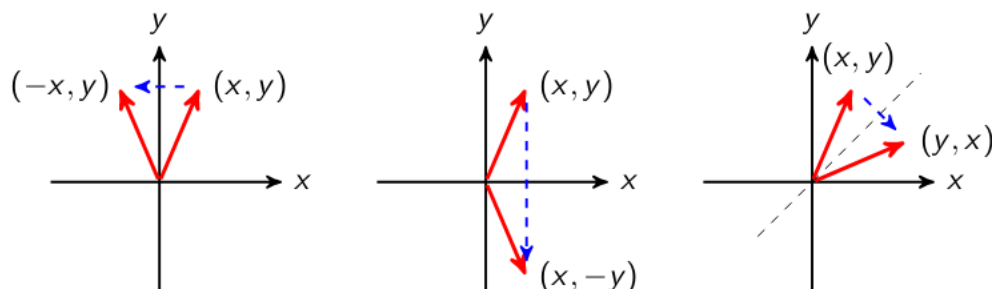
Consider linear operators f_A on \mathbb{R}^2 where A is one of the following matrices

$$\begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}, \quad \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}, \quad \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

The corresponding linear maps f_A satisfy

$$f_A(x, y) = (-x, y), f_A(x, y) = (x, -y), f_A(x, y) = (y, x), \text{ respectively}$$

They correspond to **reflections** of \mathbb{R}^2 about the y -axis, x -axis, and line $x = y$ respectively



3.2 Orthogonal Projection

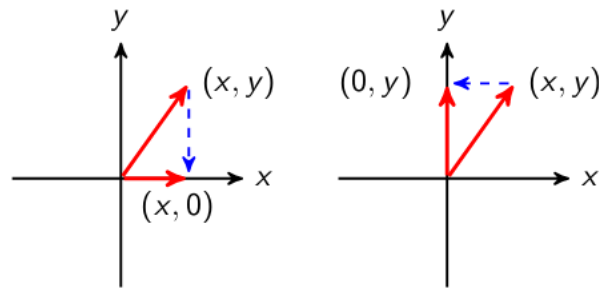
Consider linear operators f_A on \mathbb{R}^2 where A is one of the following matrices

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \quad \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

The corresponding linear maps f_A satisfy

$$f_A(x, y) = (x, 0) \text{ and } f_A(x, y) = (0, y), \text{ respectively.}$$

They correspond to the orthogonal projections of \mathbb{R}^2 onto x-axis and y-axis respectively



3.3 Rotation

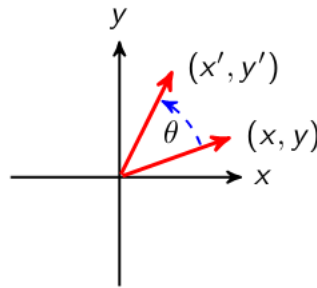
Consider the linear operator f_A on \mathbb{R}^2 where A is the following matrix:

$$\begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

The corresponding linear map f_A satisfies

$$f_A(x, y) = (x', y') = (x \cos \theta - y \sin \theta, x \sin \theta + y \cos \theta)$$

This corresponds to the rotation of \mathbb{R}^2 by angle of θ counter clock-wise



3.4 Contraction/Dilation

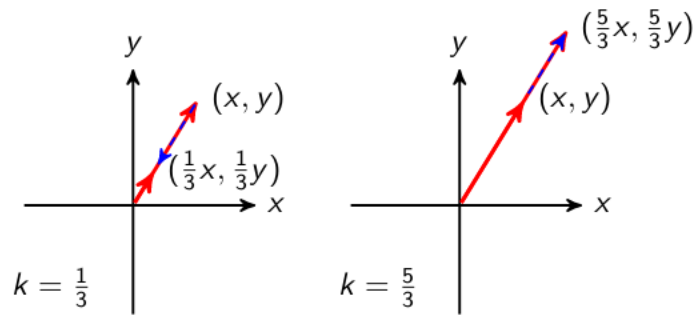
Consider linear operators f_A on \mathbb{R}^2 where A is the following matrix

$$\begin{pmatrix} k & 0 \\ 0 & k \end{pmatrix}$$

The corresponding linear map f_A satisfies

$$f_A(x, y) = (kx, ky)$$

This is **contraction** (if $0 < k < 1$) or **dilation** (if $k > 1$) of \mathbb{R}^2



3.5 Compression/Expansion

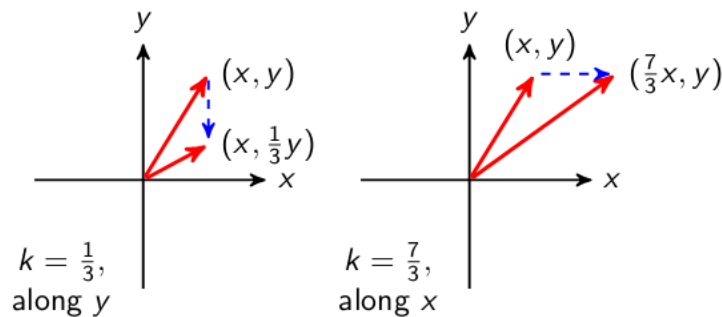
Consider linear operators f_A on \mathbb{R}^2 where A is one of the following matrices:

$$\begin{pmatrix} k & 0 \\ 0 & 1 \end{pmatrix}, \quad \begin{pmatrix} 1 & 0 \\ 0 & k \end{pmatrix}$$

The corresponding linear maps f_A satisfy

$$f_A(x, y) = (kx, y) \text{ and } f_A(x, y) = (x, ky), \text{ respectively.}$$

They correspond to **compressions** (if $0 < k < 1$) and **expansions** (if $k > 1$) of \mathbb{R}^2 along x-axis and y-axis respectively



4 Bases and Linear Maps

4.1 Theorem

Let $f : V \rightarrow W$ be a linear map where V is finite dimensional. If $S = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is a basis for V then the image of any vector $v \in V$ can be expressed as

$$f(\mathbf{v}) = c_1 f(\mathbf{v}_1) + c_2 f(\mathbf{v}_2) + \dots + c_n f(\mathbf{v}_n)$$

where c_1, \dots, c_n are the coordinates of v relative to S

4.2 Proof

Express $v \in V$ as $v = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$ and use the linearity of f

4.3 Theorem

Conversely, if $f_0 : S \rightarrow W$ is any map then the map $f : V \rightarrow W$ defined by

$$f(\mathbf{v}) = c_1 f_0(\mathbf{v}_1) + c_2 f_0(\mathbf{v}_2) + \dots + c_n f_0(\mathbf{v}_n)$$

where c_1, \dots, c_n are the coordinates of v relative to S , is a linear map

5 Exercise

Consider the basis $S = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ in \mathbb{R}^3 where $v_1 = (1, 1, 1), v_2 = (1, 1, 0), v_3 = (1, 0, 0)$ let $f : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ be a linear map such that

$$f(\mathbf{v}_1) = (1, 0), f(\mathbf{v}_2) = (2, -1), f(\mathbf{v}_3) = (4, 3)$$

Find a formula for $f(x_1, x_2, x_3)$ and use it to decide whether $f(2, -3, 5) = (9, 23)$

5.1 Solution

First express $x = (x_1, x_2, x_3)$ as $x = c_1 v_1 + c_2 v_2 + c_3 v_3$ From this we get

$$c_1 + c_2 + c_3 = x_1$$

$$c_1 + c_2 = x_2$$

$$c_1 = x_3$$

Which yields $c_1 = x_3, c_2 = x_2 - x_3, c_3 = x_1 - x_2$, so

$$\mathbf{x} = (x_1, x_2, x_3) = x_3 \mathbf{v}_1 + (x_2 - x_3) \mathbf{v}_2 + (x_1 - x_2) \mathbf{v}_3$$

Hence

$$f(x) = x_3 f(v_1) + (x_2 - x_3) f(v_2) + (x_1 - x_2) f(v_3) = (4x_1 - 2x_2 - x_3, 3x_1 - 4x_2 + x_3)$$

6 The matrix of a linear map

- Let $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear map
- Let A be the $m \times n$ matrix $[f(\mathbf{e}_1) | f(\mathbf{e}_2) | \dots | f(\mathbf{e}_n)]$ whose columns are vectors $f(e_i) \in \mathbb{R}^m$. For example, if $f : \mathbb{R}^3 \rightarrow \mathbb{R}^2$ is linear and $f(1, 0, 0) = (2, 3)$

$$f(0, 1, 0) = (0, 0), f(0, 0, 1) = (-1, 1) \text{ then } A = \begin{pmatrix} 2 & 0 & -1 \\ 3 & 0 & 1 \end{pmatrix}$$

- Note that $f(e_i) = Ae_i = f_A(e_i)$ for all i , For example

$$f(\mathbf{e}_2) = \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} = A\mathbf{e}_2 = f_A(\mathbf{e}_2)$$

- Since f and f_A agree on all vectors in a basis, we have $f = f_A$
- Hence, every linear map $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ of the form f_A for some matrix A
- This matrix A is called the (standard) matrix of linear map f
- Thus, linear maps from \mathbb{R}^n to \mathbb{R}^m are in 1-to-1 correspondence with $m \times n$ matrices (The same works for any pair of finite dimensional spaces)

7 Exercise

Find the standard matrix of the linear map $f : \mathbb{R}^4 \rightarrow \mathbb{R}^3$ defined by

$$f(x_1, x_2, x_3, x_4) = (7x_1 + 2x_2 - x_3 + x_4, x_2 + x_3, -x_1)$$

7.1 Solution

We have

$$f(1, 0, 0, 0) = (7, 0, -1)$$

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

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

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

Hence the matrix is

$$\begin{pmatrix} 7 & 2 & -1 & 1 \\ 0 & 1 & 1 & 0 \\ -1 & 0 & 0 & 0 \end{pmatrix}$$

8 The kernel and range of a linear map

8.1 Definition

Let $f : V \rightarrow W$ be a linear map

The **kernel** of f , denoted by $\ker(f)$ is defined by $\ker(f) = \{x \in V \mid f(x) = 0\}$

The **range** of f is defined as $\text{range}(f) = \{u \in W \mid u = f(x) \text{ for some } x \in V\}$

- Let A be the standard matrix of a linear map $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ (so $f(x) = Ax$)
- The $\ker(f)$ is the null space of A and $\text{range}(f)$ is the column space of A
- Use algorithms for null space and column space to find $\ker(f)$ and $\text{range}(f)$

9 Dimension theorems for matrices and linear maps

9.1 Definition

The **rank** of a linear map, denoted by $\text{rank}(f)$, is the dimension of $\text{range}(f)$

The **nullity** of f , denoted by $\text{nullity}(f)$, is the dimension of $\ker(f)$

Recall that, for a matrix A , $\text{rank}(A)$ and $\text{nullity}(A)$ are the dimensions of the column space and the null space of A . If A is the standard matrix of f then $\text{rank}(A) = \text{rank}(f)$ and $\text{nullity}(f) = \text{nullity}(A)$

9.2 Theorem (Dimension theorem for Matrices)

For any matrix A with n columns, $\text{rank}(A) + \text{nullity}(A) = n$

9.3 Theorem (Dimension Theorem for Linear Maps)

If f is a linear map from \mathbb{R}^n to \mathbb{R}^m then $\text{rank}(f) + \text{nullity}(f) = n$

10 Exercise

If $f : \mathbb{R}^5 \rightarrow \mathbb{R}^3$ is a linear map, what are the possible pairs $(\text{rank}(f), \text{nullity}(f))$?

10.1 Solution

$(0, 5), (1, 4), (2, 3), (3, 2)$. The pairs $(4, 1)$ and $(5, 0)$ are not possible because \mathbb{R}^3 does not have a subspace of dimension > 3