

## Math 270 Linear Algebra

### Chapter 6 Linear Transformations and Matrices

#### 6.1 Definitions and Examples

**Definition** Let  $V$  and  $W$  be vector spaces. A function  $L : V \rightarrow W$  is called a **linear transformation** of  $V$  into  $W$  if

$$(a) \quad L(\vec{u} + \vec{v}) = L(\vec{u}) + L(\vec{v}) \text{ for } \vec{u}, \vec{v} \in V$$

$$(b) \quad L(c\vec{u}) = cL(\vec{u}) \text{ for } \vec{u} \in V, c \in \mathbb{R}$$

If  $V = W$ , then  $L : V \rightarrow V$  is called a **linear operator** on  $V$ .

**Note:**

(1)  $+$  on the left is  $+$  in  $V$ ;  $+$  on the right is  $+$  in  $W$ .

(2)  $c\vec{u}$  on the left is scalar product in  $V$ ;  $cL(\vec{u})$  on the right is scalar product in  $W$ .

(3)  $L$  is a linear transformation  $\Leftrightarrow L(a\vec{u} + b\vec{v}) = aL(\vec{u}) + bL(\vec{v})$ . (Exercise 6)

**Example 1** Let  $A$  be  $m \times n$ .

Define  $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$  by

$$L(\vec{u}) = A\vec{u}$$

Show that  $L$  is a linear transformation.

Solution:

(a) For  $\vec{u}, \vec{v} \in \mathbb{R}^n$

$$L(\vec{u} + \vec{v}) = A(\vec{u} + \vec{v}) = A\vec{u} + A\vec{v} = L(\vec{u}) + L(\vec{v})$$

(b) For  $c$  scalar,  $\vec{u} \in \mathbb{R}^n$

$$L(c\vec{u}) = A(c\vec{u}) = cA\vec{u} = cL(\vec{u})$$

**Some important matrix transformations** (from Section 1.6)

1. reflection wrt the  $x$ -axis

$$L : \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

$$L\left(\begin{bmatrix} u_1 \\ u_2 \end{bmatrix}\right) = \begin{bmatrix} u_1 \\ -u_2 \end{bmatrix}$$

2. projection into the  $xy$ -plane

$$L : \mathbb{R}^3 \rightarrow \mathbb{R}^2$$

$$L\left(\begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}\right) = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

3. dilation

$$L : \mathbb{R}^3 \rightarrow \mathbb{R}^3$$

$$L(\vec{u}) = r\vec{u} \text{ for } r > 1$$

## 4. contraction

$$L: \mathbb{R}^3 \rightarrow \mathbb{R}^3$$

$$L(\vec{u}) = r\vec{u} \text{ for } 0 < r < 1$$

5. counterclockwise rotation through an angle of  $\phi$ 

$$L: \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

$$L(\vec{u}) = \begin{bmatrix} \cos \phi & -\sin \phi \\ \sin \phi & \cos \phi \end{bmatrix} \vec{u}$$

**Exercise #2a** Determine whether  $L$  is a linear transformation or not.

$$L: \mathbb{R}_3 \rightarrow \mathbb{R}_3$$

$$L([u_1 \ u_2 \ u_3]) = [u_1 \ u_2^2 + u_3^2 \ u_3^2]$$

Solution:

$$\text{Let } \vec{u} = [u_1 \ u_2 \ u_3], \vec{v} = [v_1 \ v_2 \ v_3] \in \mathbb{R}_3$$

$$\begin{aligned} L(\vec{u} + \vec{v}) &= L([u_1 \ u_2 \ u_3] + [v_1 \ v_2 \ v_3]) \\ &= L([u_1 + v_1 \ u_2 + v_2 \ u_3 + v_3]) \\ &= [u_1 + v_1 \ (u_2 + v_2)^2 + (u_3 + v_3)^2 \ (u_3 + v_3)^2] \end{aligned}$$

$$\begin{aligned} L(\vec{u}) + L(\vec{v}) &= [u_1 \ u_2^2 + u_3^2 \ u_3^2] + [v_1 \ v_2^2 + v_3^2 \ v_3^2] \\ &= [u_1 + v_1 \ u_2^2 + u_3^2 + v_2^2 + v_3^2 \ u_3^2 + v_3^2] \end{aligned}$$

$$\Rightarrow L(\vec{u} + \vec{v}) \neq L(\vec{u}) + L(\vec{v})$$

Thus,  $L$  is NOT a linear transformation.

**Example 5 p. 366**

$W$  = set of all real-valued functions

$V$  = subspace of all differentiable functions

$L: V \rightarrow W$  is defined by  $L(f) = f'$ . Show  $L$  is a linear transformation. (Exercise 24)

Proof:

(a) Let  $f, g \in V$

$$L(f + g) = (f + g)' = f' + g' = L(f) + L(g)$$

(b) For  $c \in \mathbb{R}, f \in V$

$$L(cf) = (cf)' = cf' = cL(f)$$

Thus,  $L$  is a linear transformation.

**Theorem** If  $L : V \rightarrow W$  is a linear transformation, then

$$(a) L(\vec{0}_V) = \vec{0}_W$$

$$(b) L(\vec{u} - \vec{v}) = L(\vec{u}) - L(\vec{v}), \text{ for } \vec{u}, \vec{v} \in V$$

Proof: (watch the video)

We can describe what a linear transformation does for all the vectors in  $V$  by looking at what it does for a basis for  $V$ :

**Theorem** Let  $L : V \rightarrow W$  be a linear transformation of an  $n$ -dimensional vector space  $V$  into a vector space  $W$ . Let  $S = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$  be a basis for  $V$ . If  $\vec{v}$  is any vector in  $V$ , then

$$L(\vec{v}) \text{ is completely determined by } \{L(\vec{v}_1), L(\vec{v}_2), \dots, L(\vec{v}_n)\}.$$

Proof:

$$\vec{v} \in V \Rightarrow \vec{v} = a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_n \vec{v}_n, \text{ where } a_i \text{ are real and unique}$$

$$\Rightarrow L(\vec{v}) = L(a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_n \vec{v}_n) = a_1 L(\vec{v}_1) + a_2 L(\vec{v}_2) + \dots + a_n L(\vec{v}_n)$$

$$\Rightarrow L(\vec{v}) \text{ is completely determined by } \{L(\vec{v}_1), L(\vec{v}_2), \dots, L(\vec{v}_n)\}$$

### Restatement of the Theorem

Let  $L : V \rightarrow W$  and  $L' : V \rightarrow W$  be linear transformations of an  $n$ -dimensional vector space  $V$  into  $W$ . Let  $S = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$  be a basis for  $V$ .

If  $L'(\vec{v}_i) = L(\vec{v}_i)$  for  $i = 1, 2, \dots, n$  then  $L'(\vec{v}) = L(\vec{v})$  for every  $\vec{v} \in V$ , i.e. if  $L$  and  $L'$  agree on a basis for  $V$  then  $L$  and  $L'$  are identical linear transformations.

**Exercise #12**  $L : \mathbb{R}_2 \rightarrow \mathbb{R}_2$ ,  $L\left(\begin{bmatrix} 1 \\ 1 \end{bmatrix}\right) = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$ ,  $L\left(\begin{bmatrix} -1 \\ 1 \end{bmatrix}\right) = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$

Find  $L\left(\begin{bmatrix} -1 \\ 5 \end{bmatrix}\right)$ .

Solution:

$$\begin{bmatrix} -1 \\ 5 \end{bmatrix} = a \begin{bmatrix} 1 \\ 1 \end{bmatrix} + b \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\Rightarrow \begin{cases} a - b = -1 \\ a + b = 5 \end{cases} \Rightarrow a = 2, b = 3$$

$$\begin{bmatrix} -1 \\ 5 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 3 \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$L\left(\begin{bmatrix} -1 \\ 5 \end{bmatrix}\right) = 2L\left(\begin{bmatrix} 1 \\ 1 \end{bmatrix}\right) + 3L\left(\begin{bmatrix} -1 \\ 1 \end{bmatrix}\right)$$

$$= 2 \begin{bmatrix} 1 \\ -2 \end{bmatrix} + 3 \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

$$= \begin{bmatrix} 2 \\ -4 \end{bmatrix} + \begin{bmatrix} 6 \\ 9 \end{bmatrix}$$

$$= \begin{bmatrix} 8 \\ 5 \end{bmatrix}$$

Next, to show that if  $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a linear transformation then  $L$  must be a matrix transformation.

**Theorem** Let  $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear transformation and  $\{\bar{e}_1, \bar{e}_2, \dots, \bar{e}_n\}$  be the natural basis for  $\mathbb{R}^n$ . Let  $A$  be the matrix whose  $j$ th column is  $L(\bar{e}_j)$ . If  $\bar{x} = [x_1 \ x_2 \ \dots \ x_n]^T \in \mathbb{R}^n$ , then

$$L(\bar{x}) = A\bar{x}. \quad (1)$$

Moreover,  $A$  is the only matrix satisfying (1) and  $A$  is called the **standard matrix representing  $L$** .

Proof:

$$\begin{aligned} \bar{x} &= x_1 \bar{e}_1 + x_2 \bar{e}_2 + \dots + x_n \bar{e}_n \\ L(\bar{x}) &= x_1 L(\bar{e}_1) + x_2 L(\bar{e}_2) + \dots + x_n L(\bar{e}_n) \\ L(\bar{x}) &= A\bar{x} \end{aligned}$$

**Exercise 35:** Show  $A$  is unique.

**Exercise #6** Find  $A$ .

a)  $L : \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$$L\left(\begin{bmatrix} u_1 \\ u_2 \end{bmatrix}\right) = \begin{bmatrix} -u_2 \\ -u_1 \end{bmatrix}$$

Solution:

$$\left. \begin{aligned} L\left(\begin{bmatrix} 1 \\ 0 \end{bmatrix}\right) &= \begin{bmatrix} 0 \\ -1 \end{bmatrix} \\ L\left(\begin{bmatrix} 0 \\ 1 \end{bmatrix}\right) &= \begin{bmatrix} -1 \\ 0 \end{bmatrix} \end{aligned} \right\} \Rightarrow A = \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$$

Thus,

$$L\left(\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right) = \begin{bmatrix} -2 \\ -1 \end{bmatrix} \text{ by definition}$$

$$\text{or } L\left(\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right) = \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} -2 \\ -1 \end{bmatrix} \text{ by using } A$$

b)  $L : \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$$\begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \rightarrow \begin{bmatrix} u_1 + ku_2 \\ u_2 \end{bmatrix}$$

Solution:

$$\begin{aligned} \begin{bmatrix} 1 \\ 0 \end{bmatrix} &\rightarrow \begin{bmatrix} 1+k \cdot 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\ \begin{bmatrix} 0 \\ 1 \end{bmatrix} &\rightarrow \begin{bmatrix} 0+k \cdot 1 \\ 1 \end{bmatrix} = \begin{bmatrix} k \\ 1 \end{bmatrix} \end{aligned}$$

Thus,  $A = \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}$ .