# CHAPTER

# 6

# **Linear Transformations**

# LEARNING OBJECTIVES

After studying the material in this chapter, you should be able to :

- Know a special class of functions, known as linear transformations.
- Understand elementary properties of linear transformations.
- Find a linear transformation by knowing its action an a basis.
- Find the matrix of a linear transformation.
- Know the Dimension Theorem, exhibiting an important relationship between the dimensions of the domain and the range of a linear transformation.
- Identify two special types of linear transformations : one-to-one and onto.
- Determine whether two vector spaces are isomorphic.
- Represent each two-dimensional point by a corresponding set of homogeneous coordinates.
- Represent all possible movements using matrix multiplication in homogeneous coordinates.
- Use Similarity Method to perform movements that are not centered at the origin.
- *Find the matrix for any composition of translations, rotations, reflections and scaling.*

## 6.1 INTRODUCTION TO LINEAR TRANSFORMATIONS

In this section we introduce a special class of functions, known as linear transformations, that map vectors in one vector space to those in another. We will also examine some elementary properties of linear transformations.

### **DEFINITION** Linear Transformation

Let V and W be two vector spaces over  $\mathbb{R}$ . A function

$$T : V \to W$$

is called a **linear transformation** from V to W if it satisfies the following properties:

1.  $T(v_1 + v_2) = T(v_1) + T(v_2)$ , for all  $v_1, v_2 \in V$ 

**2.**  $T(\alpha v) = \alpha T(v)$ , for all  $\alpha \in \mathbb{R}$  and all  $v \in V$ .

Thus, a linear transformation is a function from one vector space to another that preserves the operations of addition and scalar multiplication.

Note Notice that the two conditions for linearity are equivalent to a single condition

$$T(\alpha v_1 + \beta v_2) = \alpha T(v_1) + \beta T(v_2)$$
, for all  $v_1, v_2 \in V$  and all  $\alpha, \beta \in \mathbb{R}$ .

#### **EXAMPLE 1** Zero Linear Transformation

Let *V* and *W* be vector spaces. Consider the mapping  $T: V \to W$  defined by  $T(v) = \mathbf{0}_W$ , for all  $v \in V$ . We will show that *T* is a linear transformation.

- 1. We must show that  $T(v_1 + v_2) = T(v_1) + T(v_2)$ , for all  $v_1, v_2 \in V$ Now  $T(v_1 + v_2) = \mathbf{0}_W = \mathbf{0}_W + \mathbf{0}_W = T(v_1) + T(v_2)$ .
- 2. We must show that  $T(\alpha v) = \alpha T(v)$ , for all  $\alpha \in \mathbb{R}$  and for all  $v \in V$ Now  $T(\alpha v) = \mathbf{0}_W = \alpha \mathbf{0}_W = \alpha T(v)$ .

Hence, T is a linear transformation, known as the zero linear transformation.

#### EXAMPLE 2 Let

and

 $V = \mathcal{M}_{mn}$ , the space of all  $m \times n$  matrices  $W = \mathcal{M}_{nm}$ , the space of all  $n \times m$  matrices

anu

Consider the mapping  $T: V \rightarrow W$  defined by

 $T(A) = A^T$  for all  $A \in V$ 

Show that *T* is a linear transformation.

**SOLUTION** Let  $A_1$  and  $A_2$  be any two matrices in  $V = \mathcal{M}_{mn}$ .

Then

$$T(A_1 + A_2) = (A_1 + A_2)^T = A_1^T + A_2^T = T(A_1) + T(A_2)$$

Similarly,

$$T(\alpha A) = (\alpha A)^T = \alpha A^T = \alpha T(A)$$
, for any  $\alpha \in \mathbb{R}$  and  $A \in V$ 

Hence, T is a linear transformation from  $\mathcal{M}_{mn}$  to  $\mathcal{M}_{nm}$ .

#### EXAMPLE 3 Let

 $V = \mathcal{P}_n$ , the space of all polynomials of degree  $\leq n$ , with real coefficients and  $W = \mathcal{P}_{n-1}$ , the space of all polynomials of degree  $\leq n-1$ , with real coefficients Consider the mapping  $T: V \to W$  defined by

$$T(\mathbf{p}) = \mathbf{p}' \text{ for any } \mathbf{p} \in \mathbf{V} = \mathbf{P}_n$$

Show that *T* is a linear transformation.

**SOLUTION** For any  $p_1, p_2 \in V$ , we have

$$T(p_1 + p_2) = (p_1 + p_2)' = p_1' + p_2' = T(p_1) + T(p_2)$$

Similarly,

$$T(\alpha p) = (\alpha p)' = \alpha p' = \alpha T(p)$$
, for any  $\alpha \in \mathbb{R}$  and  $p \in V$ 

Hence, T is a linear transformation.

**EXAMPLE 4** Let *V* be an *n*-dimensional vector space over  $\mathbb{R}$ , and let *B* be an ordered basis for *V*. Then every vector  $\mathbf{v} \in V$  has its coordinatization  $[\mathbf{v}]_B$  with respect to *B* satisfying the following properties

 $[v_1 + v_2]_B = [v_1]_B + [v_2]_B$ , for all  $v_1, v_2 \in V$ 

 $[\alpha v]_{R} = \alpha [v]_{R}$ , for all  $\alpha \in \mathbb{R}$ , and for all  $v \in V$ 

Consider the mapping  $T: V \to \mathbb{R}^n$  defined by

$$T(\mathbf{v}) = [\mathbf{v}]_{R}$$
 for any  $\mathbf{v} \in V$ 

We will show that T is a linear transformation. Let  $v_1$  and  $v_2$  be any two vectors in V. Then from the properties of coordinatization just stated, we have

$$T(v_1 + v_2) = [v_1 + v_2]_B = [v_1]_B + [v_2]_B = T(v_1) + T(v_2)$$

Similarly,

$$T(\alpha \mathbf{v}) = [\alpha \mathbf{v}]_B = \alpha [\mathbf{v}]_B = \alpha T(\mathbf{v}), \text{ for any } \alpha \in \mathbb{R} \text{ and } \mathbf{v} \in V$$

Hence, T is a linear transformation.

#### **DEFINITION** Linear Operator

Let V be a vector space. A linear transformation  $T: V \rightarrow V$  is called a **linear operator**. Thus, a linear operator is a linear transformation from a vector space to itself.

#### **EXAMPLE 5** Identity Linear Operator

Let V be a vector space. Consider the mapping  $T: V \to V$  defined by T(v) = v for all  $v \in V$ . We will show that T is a linear operator. Let  $v_1, v_2 \in V$ . Then

 $T(v_1 + v_2) = v_1 + v_2 = T(v_1) + T(v_2)$ 

Also, let  $v \in V$  and  $\alpha \in \mathbb{R}$ . Then

 $T(\alpha \mathbf{v}) = \alpha \mathbf{v} = \alpha T(\mathbf{v})$ 

Hence, T is a linear operator, known as the Identity Linear Operator.

#### **EXAMPLE 6** Contractions and Dilations

Let  $k \in \mathbb{R}$ . Define  $T : \mathbb{R}^n \to \mathbb{R}^n$  as  $T(\mathbf{v}) = k\mathbf{v}$ , for all  $\mathbf{v} \in \mathbb{R}^n$ .

We will show that T is a linear operator.

1. Let  $v_1, v_2 \in \mathbb{R}^n$ . Then

$$T(v_1 + v_2) = k(v_1 + v_2) = kv_1 + kv_2 = T(v_1) + T(v_2)$$

**2.** Let  $v \in \mathbb{R}^n$  and  $\alpha \in \mathbb{R}$ . Then

$$T(\alpha v) = k(\alpha v) = \alpha(kv) = \alpha T(v)$$

Hence, T is a linear operator, called **dilation** or **contraction**, according as |k| > 1 or |k| < 1, respectively. If |k| > 1, T **dilates** (stretches) the length of the vector, and if |k| < 1, T **contracts** (shrinks) the length.

#### **EXAMPLE 7** Projections

Consider the mapping  $T: \mathbb{R}^3 \to \mathbb{R}^3$  defined by

 $T([x_1, x_2, x_3]) = [x_1, x_2, 0], \quad [x_1, x_2, x_3] \in \mathbb{R}^3$ 

We will show that *T* is a linear operator.

1. Let 
$$\mathbf{v_1} = [x_1, x_2, x_3], \ \mathbf{v_2} = [y_1, y_2, y_3] \in \mathbb{R}^3$$
. Then  
 $T(\mathbf{v_1} + \mathbf{v_2}) = T([x_1, x_2, x_3] + [y_1, y_2, y_3])$   
 $= T([x_1 + y_1, x_2 + y_2, x_3 + y_3])$   
 $= [x_1 + y_1, x_2 + y_2, 0]$   
 $= [x_1, x_2, 0] + [y_1, y_2, 0]$   
 $= T([x_1, x_2, x_3]) + T([y_1, y_2, y_3])$   
 $= T(\mathbf{v_1}) + T(\mathbf{v_2})$ 

2. Let 
$$\mathbf{v} = [x_1, x_2, x_3] \in \mathbb{R}^3$$
 and  $\alpha \in \mathbb{R}$ . Then  
 $T(\alpha \, \mathbf{v}) = T(\alpha[x_1, x_2, x_3])$   
 $= T([\alpha x_1, \alpha x_2, \alpha x_3])$   
 $= [\alpha x_1, \alpha x_2, 0]$   
 $= \alpha T([x_1, x_2, x_3])$   
 $= \alpha T([\mathbf{v})$ 

Hence, *T* is a linear operator on  $\mathbb{R}^3$ , known as a **projection operator**, because of its geometrical interpretation. *It projects each vector in*  $\mathbb{R}^3$  *to a corresponding vector in the xy-plane* (see Fig. 6.1).



FIGURE 6.1

**Note** Notice that we can also define a projection operator on  $\mathbb{R}^3$  which projects each vector in  $\mathbb{R}^3$  to a corresponding vector in the *yz*-plane or the *zx*-plane.

#### **EXAMPLE 8** Reflections

Consider the mapping  $T: \mathbb{R}^3 \to \mathbb{R}^3$  defined by

 $T([x_1, x_2, x_3]) = [x_1, x_2, -x_3], \quad [x_1, x_2, x_3] \in \mathbb{R}^3$ 

We will show that *T* is a linear operator.

1. Let 
$$\mathbf{v_1} = [x_1, x_2, x_3], \ \mathbf{v_2} = [y_1, y_2, y_3] \in \mathbb{R}^3$$
. Then  
 $T(\mathbf{v_1} + \mathbf{v_2}) = T([x_1, x_2, x_3] + [y_1, y_2, y_3])$   
 $= T([x_1 + y_1, x_2 + y_2, x_3 + y_3])$   
 $= [x_1 + y_1, x_2 + y_2, -(x_3 + y_3)]$   
 $= [x_1, x_2, -x_3] + [y_1, y_2, -y_3]$   
 $= T([x_1, x_2, x_3]) + T([y_1, y_2, y_3])$   
 $= T(\mathbf{v_1}) + T(\mathbf{v_2})$ 

2. Let 
$$\mathbf{v} = [x_1, x_2, x_3] \in \mathbb{R}^3$$
 and  $\alpha \in \mathbb{R}$ . Then  
 $T(\alpha \, \mathbf{v}) = T(\alpha [x_1, x_2, x_3])$   
 $= T([\alpha x_1, \alpha x_2, \alpha x_3])$   
 $= [\alpha x_1, \alpha x_2, -\alpha x_3]$   
 $= \alpha [x_1, x_2, -\alpha x_3]$   
 $= \alpha T([x_1, x_2, x_3])$   
 $= \alpha T([\mathbf{v})$ 

Hence, *T* is a linear operator on  $\mathbb{R}^3$ , called a **reflection operator**. This operator reflects each vector  $[x_1, x_2, x_3]$  through the *xy*-plane, which acts like a mirror (see Fig. 6.2).



FIGURE 6.2

**Note** Notice that we can also define a reflection operator on  $\mathbb{R}^3$  which reflects each vector in  $\mathbb{R}^3$  through the *yz*-plane or the *zx*-plane.

## **EXAMPLE 9** Rotation Linear Operator

Let  $T: \mathbb{R}^2 \to \mathbb{R}^2$  be defined by

$$T(\mathbf{v}) = T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x\cos\theta - y\sin\theta \\ x\sin\theta + y\cos\theta \end{bmatrix},$$

for all  $v = [x, y] \in \mathbb{R}^2$ , where  $\theta$  is fixed angle. We will show that *T* is a linear operator.

Let  $\mathbf{v_1} = [x_1, y_1], \mathbf{v_2} = [x_2, y_2]$  be two vectors in  $\mathbb{R}^2$ . Then

$$T(\mathbf{v}_{1} + \mathbf{v}_{2}) = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} (\mathbf{v}_{1} + \mathbf{v}_{2})$$
$$= \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} (\mathbf{v}_{1}) + \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} (\mathbf{v}_{2})$$
$$= T(\mathbf{v}_{1}) + T(\mathbf{v}_{2})$$

Next, let  $\alpha \in \mathbb{R}$  and  $v = [x, y] \in \mathbb{R}^2$ . Then

$$T(\alpha v) = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} (\alpha v) = \alpha \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} v = \alpha T(v)$$

Hence *T* is a linear operator.

**Note** Notice that the linear operator T defined above rotates the vector [x, y] counterclockwise through the angle  $\theta$  in the plane (see Figure 6.3). To prove this, consider the vector [x', y']obtained by rotating [x, y] counterclockwise through the angle  $\theta$ . We can write  $x = r \cos \alpha$ , y = r

 $\sin \alpha$ , where  $r = \sqrt{x^2 + y^2}$ , and  $\alpha$  is the angle shown in Figure 6.3.



#### **FIGURE 6.3**

Also,  $x' = r \cos(\theta + \alpha)$ , and  $y' = r \sin(\theta + \alpha)$ Using the following trigonometry formulas:

 $a(A \perp D)$ 

$$cos(A + B) = cos A cos B - sin A sin B$$
  

$$sin(A + B) = sin A cos B + cos A sin B$$

we see that

$$x' = r\cos(\theta + \alpha) = r\cos\theta\cos\alpha - r\sin\theta\sin\alpha = x\cos\theta - y\sin\theta$$
$$y' = r\sin(\theta + \alpha) = r\sin\theta\cos\alpha + r\cos\theta\sin\alpha = x\sin\theta + y\cos\theta$$

$$\therefore \qquad \begin{bmatrix} x'\\ y' \end{bmatrix} = \begin{bmatrix} x\cos\theta - y\sin\theta\\ x\sin\theta + y\cos\theta \end{bmatrix} = \begin{bmatrix} \cos\theta & -\sin\theta\\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x\\ y \end{bmatrix} = T\left(\begin{bmatrix} x\\ y \end{bmatrix}\right)$$
  
*i.e.*, 
$$T\left(\begin{bmatrix} x\\ y \end{bmatrix}\right) = \begin{bmatrix} x'\\ y' \end{bmatrix}.$$

**EXAMPLE 10** Let *A* be a fixed  $m \times n$  matrix, and let  $T : \mathbb{R}^n \to \mathbb{R}^m$  defined by

$$T(\mathbf{x}) = A\mathbf{x}$$
, for all  $\mathbf{x} \in \mathbb{R}^n$ 

Show that *T* is a linear transformation.

**SOLUTION** Let  $x_1, x_2 \in \mathbb{R}^n$ . Then  $T(x_1 + x_2) = A(x_1 + x_2) = Ax_1 + Ax_2 = Tx_1 + Tx_2$ Also, let  $x \in \mathbb{R}^n$  and  $c \in \mathbb{R}$ . Then T(cx) = A(cx) = c(Ax) = cT(x). Hence, *T* is a linear transformation.

#### **EXAMPLE 11** Shear Operators

Let *k* be a fixed scalar in  $\mathbb{R}$ . Consider the function  $T: \mathbb{R}^2 \to \mathbb{R}^2$  defined by

$$T\left(\begin{bmatrix} x\\ y \end{bmatrix}\right) = \begin{bmatrix} 1 & k\\ 0 & 1 \end{bmatrix} \begin{bmatrix} x\\ y \end{bmatrix} = \begin{bmatrix} x+ky\\ y \end{bmatrix}$$

Show that *T* is a linear operator.

[Delhi Univ. GE-2, 2018]

**SOLUTION** Let  $v_1 = [x_1, y_1]$  and  $v_2 = [x_2, y_2]$  be two vectors in  $\mathbb{R}^2$ . Then

$$T(v_1 + v_2) = \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix} (v_1 + v_2) = \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix} v_1 + \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix} v_2 = T(v_1) + T(v_2)$$

Next, let  $\alpha \in \mathbb{R}$  and  $\mathbf{v} = [x, y] \in \mathbb{R}^2$ . Then

$$T(\alpha v) = \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix} (\alpha v) = \alpha \left( \begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix} v \right) = \alpha T(v)$$

Hence, *T* is a linear operator, called a **shear in the** *x***-axis with factor** *k*. Similarly, the function  $T : \mathbb{R}^2 \to \mathbb{R}^2$  defined by

$$T\left(\begin{bmatrix} x\\ y \end{bmatrix}\right) = \begin{bmatrix} 1 & 0\\ k & 1 \end{bmatrix} \begin{bmatrix} x\\ y \end{bmatrix} = \begin{bmatrix} x\\ kx + y \end{bmatrix}$$

is also a linear operator on  $\mathbb{R}^2$ , called a shear in the *y*-direction with factor *k*.



FIGURE 6.4

The following theorem contains some basic properties of linear transformations.

#### **THEOREM 6.1** Properties of Linear Transformations

Let *V* and *W* be two vector spaces, and let  $T: V \to W$  be a linear transformation. Let  $\mathbf{0}_V$  be the zero vector in *V* and  $\mathbf{0}_W$  be the zero vector in *W*. Then

**1.** 
$$T(\mathbf{0}_V) = \mathbf{0}_W$$

**2.**  $T(-\mathbf{v}) = -T(\mathbf{v})$ , for all  $\mathbf{v} \in V$ 

3. 
$$T(\boldsymbol{u} - \boldsymbol{v}) = T(\boldsymbol{u}) - T(\boldsymbol{v})$$
, for all  $\boldsymbol{u}, \boldsymbol{v} \in V$ 

4. 
$$T(a_1 v_1 + a_2 v_2 + ... + a_n v_n) = a_1 T(v_1) + a_2 T(v_2) + ... + a_n T(v_n)$$
, for all  $a_1, a_2, ..., a_n \in \mathbb{R}$ ,  
 $v_1, v_2, ..., v_n \in V$ , for  $n \ge 1$ 

Proof 1.  $T(\mathbf{0}_{V}) = T(\mathbf{0} \ \mathbf{0}_{V})$  $= 0 T(\mathbf{0}_{V})$ (Property (2) of linear transformation)  $= 0_{W}$ So, (1) is proved.  $T(-\mathbf{v}) = T((-1)\mathbf{v})$ 2. = (-1)T(v)(Property (2) of linear transformation)  $= -T(\mathbf{v})$ So, (2) is proved.  $T(\boldsymbol{u} - \boldsymbol{v}) = T(\boldsymbol{u} + (-1)\boldsymbol{v})$ 3.  $= T(\boldsymbol{u}) + T(-1)\boldsymbol{v}$ (Property (1) of linear transformation)  $= T(\boldsymbol{u}) - T(\boldsymbol{v})$ (by part (2)) 4. To prove (4), we use induction on n. For n = 1, we have  $T(a_1 \boldsymbol{v_1}) = a_1 T(\boldsymbol{v_1})$ (Property (2) of linear transformation) so, result is true for n = 1Similarly, for n = 2, we have  $T(a_1 v_1 + a_2 v_2) = T(a_1 v_1) + T(a_2 v_2)$ (Property (1) of linear transformation)  $= a_1 T(\mathbf{v_1}) + a_2 T(\mathbf{v_2})$ (Property (2) of linear transformation) so, the result is also true for n = 2Now, we assume that the result is true for n = m, *i.e.*,  $T(a_1v_1 + a_2v_2 + \dots + a_mv_m) = a_1T(v_1) + a_2T(v_2) + \dots + a_mT(v_m)$ We have to deduce that the result is also true for n = m + 1We have  $T(a_1 v_1 + a_2 v_2 + \dots + a_m v_m + a_{m+1} v_{m+1}) = T(a_1 v_1 + a_2 v_2 + \dots + a_m v_m) + T(a_{m+1} v_{m+1})$ 

(Property (2) of linear transformation)

$$= a_1 T(\mathbf{v_1}) + a_2 T(\mathbf{v_2}) + \dots + a_m T(\mathbf{v_m}) + a_{m+1} T(\mathbf{v_{m+1}})$$
  
(by the induction hypothesis)

So, the result is true for n = m + 1. Hence, by the principle of mathematical induction, the result is true for any natural number n.

Note Part (1) of Theorem 6.1 can be used to prove that a function is not a linear transformation.

**EXAMPLE 12** Let V be a vector space, and let  $x \neq 0$  be a fixed vector in V. Prove that the translation function  $f: V \to V$  defined by f(v) = v + x is not a linear transformation.

[Delhi Univ. GE-2, 2018]

SOLUTION We have

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

So, by part (1) of Theorem 6.1 f is not a linear transformation.

#### **Composition of Linear Transformations**

If  $f: X \to Y$  and  $g: Y \to Z$  are functions, then the **composition** of f and g is defined to be the function  $gof: X \to Z$  given by (gof)(x) = g(f(x)). The following theorem asserts that the composition of linear transformations is again a linear transformation.

**THEOREM 6.2** Let V, W and X be vector spaces. Let  $T_1: V \to W$  and  $T_2: W \to X$  be linear transformations. Then the composition function  $T_2 o T_1 : V \to X$  given by  $(T_2 o T_1) (v) = T_2 (T_1(v))$ , for all  $v \in V$ , is a linear transformation.

**Proof** To show that  $T_2 o T_1$  is a linear transformation, we must show that both of the following are true:

$$(T_2 \circ T_1)(v_1 + v_2) = (T_2 \circ T_1)(v_1) + (T_2 \circ T_1)(v_2), \text{ for all } v_1, v_2 \in V$$
  
$$(T_2 \circ T_1)(\alpha v) = \alpha (T_2 \circ T_1)(v), \text{ for all } \alpha \in \mathbb{R} \text{ and } v \in V$$

To prove the first property, consider

 $\begin{array}{l} = T_2 \left( T_1(\boldsymbol{v_1} + \boldsymbol{v_2}) \right) & (\text{definition of composition}) \\ = T_2 \left[ T_1(\boldsymbol{v_1}) + T_1(\boldsymbol{v_2}) \right] & (\because T_1 \text{ is a Linear Transformation}) \\ = T_2 \left( T_1(\boldsymbol{v_1}) \right) + T_2(T_1(\boldsymbol{v_2})) & (\because T_2 \text{ is a Linear Transformation}) \\ = \left( T_2 o T_1 \right) (\boldsymbol{v_1}) + \left( T_2 o T_1 \right) (\boldsymbol{v_2}) & (\text{definition of composition}) \end{array}$  $(T_2 o T_1)(v_1 + v_2) = T_2 (T_1(v_1 + v_2))$ 

So, the first property holds.

To prove the second property, consider

$$\begin{array}{ll} (T_2 \circ T_1)(\alpha \, \mathbf{v}) &= T_2(T_1(\alpha \, \mathbf{v})) & (\text{definition of composition}) \\ &= T_2(\alpha T_1(\mathbf{v})) & (\because T_1 \text{ is a Linear Transformation}) \\ &= \alpha(T_2(T_1(\mathbf{v})) & (\because T_2 \text{ is a Linear Transformation}) \\ &= \alpha(T_2 \circ T_1)(\mathbf{v}) & (\text{definition of composition}) \end{array}$$

So, the second property also holds.

Hence,  $T_2 o T_1$  is a linear transformation.

**EXAMPLE 13** Let  $T_1 : \mathbb{R}^2 \to \mathbb{R}^2$  be the linear operator representing the counterclockwise rotation in  $\mathbb{R}^2$  through a fixed angle  $\theta$ . That is,

$$T_{1}(\boldsymbol{v}) = T_{1}\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x \cos \theta - y \sin \theta \\ x \sin \theta + y \cos \theta \end{bmatrix},$$

where  $v = [x, y] \in \mathbb{R}^2$ . Further, let  $T_2 : \mathbb{R}^2 \to \mathbb{R}^2$  be the linear operator representing the reflection of vectors in  $\mathbb{R}^2$  through the *x*-axis. That is,

$$T_2(\mathbf{v}) = T_2\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} x \\ -y \end{bmatrix}, \quad \mathbf{v} = [x, y] \in \mathbb{R}^2$$

Because  $T_1$  and  $T_2$  are both linear transformations, Theorem 6.2 asserts that the composition  $T_2 o T_1$  of  $\vec{T_1}$  and  $\vec{T_2}$  given by

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$$(T_2 o T_1)(\mathbf{v}) = T_2 (T_1(\mathbf{v})) = T_2 \left( \begin{bmatrix} x \cos \theta - y \sin \theta \\ x \sin \theta + y \cos \theta \end{bmatrix} \right) = \begin{bmatrix} x \cos \theta - y \sin \theta \\ -x \sin \theta - y \cos \theta \end{bmatrix}$$

is also a linear transformation. Notice that  $T_2 o T_1$  represents a counterclockwise rotation of [x, y] through the angle  $\theta$  followed by a reflection through the *x*-axis (see Fig. 6.5).





#### Linear Transformations and Subspaces

We conclude this section by proving that, under a linear transformation  $T: V \rightarrow W$ , "subspaces" of V are mapped to "subspaces" of W, and vice-versa.

Let  $V_1$  and  $V_2$  be vector spaces and let  $T: V_1 \to V_2$  be a linear transformation. Given a set  $U \subseteq V_1$ , the **image** of U in  $V_2$  is defined to be the set

$$T(U) = \{T(\boldsymbol{u}) : \boldsymbol{u} \in U\}$$

Similarly, given a set  $W \subseteq V_2$ , the **pre-image** of W in  $V_1$  is defined to be the set

$$T^{-1}(W) = \{ v \in V_1 : T(v) \in W \}$$

**THEOREM 6.3** Let  $V_1$  and  $V_2$  be vector spaces, and let  $T: V_1 \rightarrow V_2$  be a linear transformation.

1. If U is a subspace of  $V_1$ , then T(U), the image of U in  $V_2$ , is a subspace of  $V_2$ .

**2.** If W is a subspace of  $V_2$ , then  $T^{-1}(W)$ , the pre-image of W in  $V_1$ , is a subspace of  $V_1$ .

**Proof** 1. Since U is a subspace of  $V_1$ ,  $\mathbf{0}_{V_1} \in U$ . By part (1) of Theorem 6.1 we have

$$\mathbf{0}_{V_2} = T(\mathbf{0}_{V_1}) \in T(U)$$

Thus, T(U) is non-empty. Hence, to show that T(U) is a subspace of  $V_2$ , we must show that T(U) is closed under addition and scalar multiplication.

First, suppose that  $w_1, w_2$  are any two vectors in T(U). Then, by definition of T(U), we have

$$w_1 = T(u_1)$$
 and  $w_2 = T(u_2)$ 

for some  $u_1, u_2 \in U$ . So,

$$w_1 + w_2 = T(u_1) + T(u_2) = T(u_1 + u_2)$$
 (: T is a L.T.)

However, since U is a subspace of  $V_1$ ,  $u_1 + u_2 \in U$ . Thus,  $w_1 + w_2 \in T(U)$ . Hence, T(U) is closed under addition.

Next, let w be any vector in T(U), and let  $\alpha$  be a scalar. We must show that  $\alpha w \in T(U)$ . By definition of T(U), w = T(u), for some  $u \in U$ . Then

$$\alpha w = \alpha T(u) = T(\alpha u) \qquad (\because T \text{ is a L.T.})$$

However, since U is a subspace of  $V_1$ ,  $\alpha \mathbf{u} \in U$ , and hence  $\alpha \mathbf{w} \in T(U)$ . Thus, T(U) is closed under scalar multiplication.

2. The pre-image of a subspace W of  $V_2$  is given by  $T^{-1}(W) = \{ v \in V_1 : T(v) \in W \}$ 

$$T(\mathbf{0}_{V_1}) = \mathbf{0}_{V_2} \in W, \text{ so } \mathbf{0}_{V_1} \in T^{-1}(W) \quad \therefore \quad T^{-1}(W) \text{ is non-empty.}$$

Also, let  $\mathbf{v}_1, \mathbf{v}_2 \in T^{-1}(W) \Rightarrow T(\mathbf{v}_1), T(\mathbf{v}_2) \in W \Rightarrow T(\mathbf{v}_1) + T(\mathbf{v}_2) \in W \Rightarrow T(\mathbf{v}_1 + \mathbf{v}_2) \in W$  $\Rightarrow \mathbf{v}_1 + \mathbf{v}_2 \in T^{-1}(W)$ 

Finally, let  $v \in T^{-1}(W)$ , and let  $\alpha \in \mathbb{R}$ . Then

$$\mathbf{v} \in T^{-1}(W) \Rightarrow T(\mathbf{v}) \in W \Rightarrow \alpha T(\mathbf{v}) \in W \Rightarrow T(\alpha \mathbf{v}) \in W \Rightarrow \alpha \mathbf{v} \in T^{-1}(W)$$
  
Hence  $T^{-1}(W)$  is a subspace of  $V_1$ .

#### **EXERCISE 6.1**

- 1. Determine which of the following functions are linear transformations.
  - (a)  $T: \mathbb{R}^2 \to \mathbb{R}^2$  given by T([x, y]) = [2x 3y, 3x + 4y]

(b) 
$$T: \mathbb{R}^3 \to \mathbb{R}^3$$
 given by  $T([x_1, x_2, x_3]) = [x_1 + 1, x_2 - 2, x_3] = [x_1, x_2, x_3] + [1, -2, 0]$ 

- (c)  $T: \mathbb{R}^2 \to \mathbb{R}$  given by  $T([x, y]) = \sqrt{x^2 + y^2}$ .
- (d)  $T: \mathbb{R}^4 \to \mathbb{R}$  given by  $T([x_1, x_2, x_3, x_4]) = |x_1|$
- (e)  $T: \mathcal{P}_2 \to \mathbb{R}$  given by  $T(a_2 x^2 + a_1 x + a_0) = a_2 + a_1 + a_0$

(f) 
$$T: \mathcal{M}_{22} \to \mathbb{R}$$
 given by  $T\left(\begin{bmatrix} a & b \\ c & d \end{bmatrix}\right) = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$ 

(g) 
$$T: \mathbb{R}^3 \to \mathbb{R}^3$$
 given by  $T([x_1, x_2, x_3]) = [e^{x_1}, \cos x_2, \sin x_3]$ 

- 2. Show that the mapping  $T : \mathbb{R}^3 \to \mathbb{R}^3$  given by  $T([x_1, x_2, x_3]) = [-x_1, x_2, x_3]$  is a linear operator.
- 3. Let x be a fixed vector in  $\mathbb{R}^n$ . Prove that the mapping  $T : \mathbb{R}^n \to \mathbb{R}$  given by  $T(y) = x \cdot y$  is a linear transformation.
- 4. (a) Show that the mapping  $T : \mathcal{M}_{nn} \to \mathcal{M}_{nn}$  given by  $T(A) = A + A^T$  is a linear operator on  $\mathcal{M}_{nn}$ .