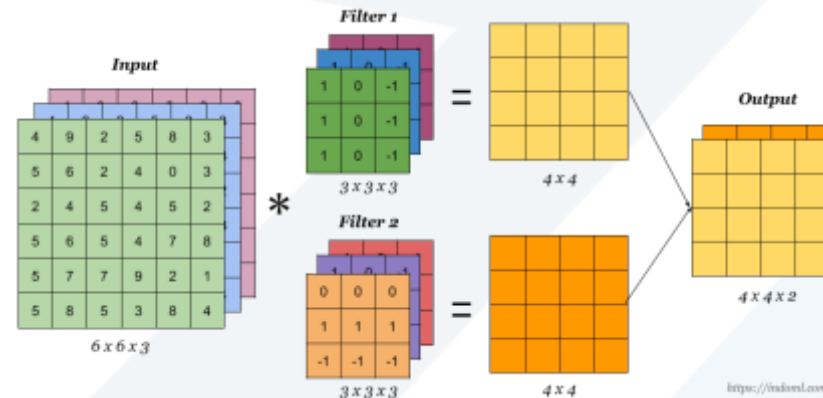


# CECC122: Linear Algebra and Matrix Theory

## Lecture Notes 7: Linear Transformations



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## Chapter 6

# Linear Transformations

1. Introduction to Linear Transformations
2. The Kernel and Range of a Linear Transformation
3. Compositions and Inverse Transformations
5. Geometry of Matrix Operators

# 1. Introduction to Linear Transformations

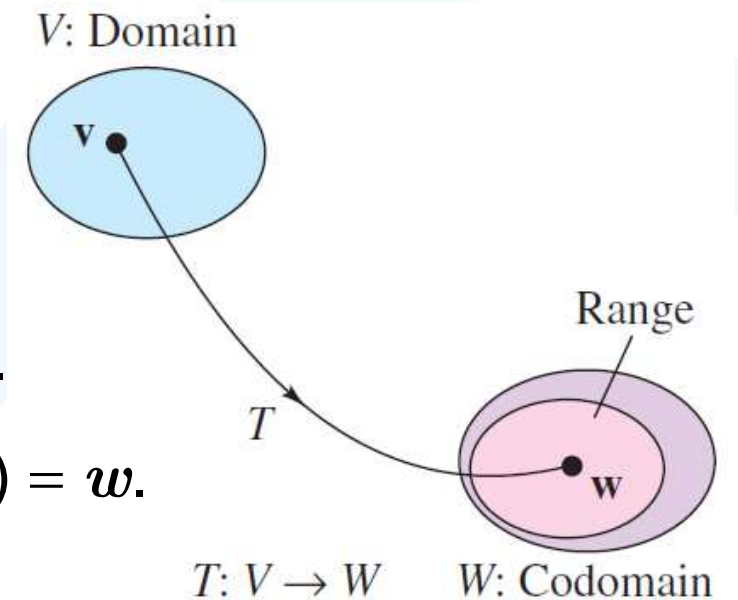
## Images And Preimages of Functions:

- Function  $T$  that maps a vector space  $V$  into a vector space  $W$ .

$$T: V \xrightarrow{\text{Mapping}} W, \quad V, W: \text{vector spaces}$$

- If  $v$  is in  $V$  and  $w$  is in  $W$  such that:  $T(v) = w$ , Then  $w$  is called the **image** of  $v$  under  $T$ .
- The range of  $T$ :** The set of all images of vectors in  $V$ .
- The preimage of  $w$ :** The set of all  $v$  in  $V$  such that  $T(v) = w$ .
- Example 1: (A function from  $R^2$  into  $R^2$ )**

$$T: R^2 \rightarrow R^2, \quad v = (v_1, v_2) \in R^2 \quad T(v_1, v_2) = (v_1 - v_2, v_1 + 2v_2)$$



(a) Find the image of  $v = (-1, 2)$       (b) Find the preimage of  $w = (-1, 11)$

$$(a) v = (-1, 2) \Rightarrow T(v) = T(-1, 2) = (-1 - 2, -1 + 2(2)) = (-3, 3)$$

$$(b) T(v) = w = (-1, 11) \Rightarrow T(v_1, v_2) = (v_1 - v_2, v_1 + 2v_2) = (-1, 11)$$

$$\Rightarrow \begin{cases} v_1 - v_2 = -1 \\ v_1 + 2v_2 = 11 \end{cases} \Rightarrow v_1 = 3, v_2 = 4$$

Thus  $\{(3, 4)\}$  is the preimage  
of  $w = (-1, 11)$

- **Definition:** If  $T : V \rightarrow W$  is a mapping from a vector space  $V$  to a vector space  $W$ , then  $T$  is called a **linear transformation (LT)** from  $V$  to  $W$  if the following two properties hold for all vectors  $u$  and  $v$  in  $V$  and for all scalars  $c$ :

$$(1) T(u + v) = T(u) + T(v) \quad \text{[Additivity property]}$$

$$(2) T(cu) = cT(u) \quad \text{[Homogeneity property]}$$

When  $V = W$ , the linear transformation  $T$  is called a **linear operator** on  $V$ .

- **Example 2: (Verifying a linear transformation  $T$  from  $R^2$  into  $R^2$ )**

$$T: R^2 \rightarrow R^2, \mathbf{v} = (v_1, v_2) \in R^2 \quad T(v_1, v_2) = (v_1 - v_2, v_1 + 2v_2)$$

$\mathbf{u} = (u_1, u_2), \mathbf{v} = (v_1, v_2)$  vectors in  $R^2, c$ : any real

$$T(\mathbf{u} + \mathbf{v}) = T(u_1 + v_1, u_2 + v_2)$$

$$= ((u_1 + v_1) - (u_2 + v_2), (u_1 + v_1) + 2(u_2 + v_2))$$

$$= ((u_1 - u_2) + (v_1 - v_2), (u_1 + 2u_2) + (v_1 + 2v_2))$$

$$= (u_1 - u_2, u_1 + 2u_2) + (v_1 - v_2, v_1 + 2v_2) = T(\mathbf{u}) + T(\mathbf{v})$$

$$T(c\mathbf{u}) = T(cu_1, cu_2) = (cu_1 - cu_2, cu_1 + 2cu_2) = c(u_1 - u_2, u_1 + 2u_2) = cT(\mathbf{u})$$

Therefore,  $T$  is a linear transformation.

- **Example 3: (A Linear Transformation from  $P_n$  to  $P_{n-1}, n \geq 1$ )**

$$T: P_n \rightarrow P_{n-1}: T(\mathbf{p}) = T(p(x)) = p'(x) \quad \text{derivative}$$

- **Example 4: (A Linear Transformation from  $P_n$  to  $P_{n+1}$ )**

$$\mathbf{p} = p(x) = c_0 + c_1x + \dots + c_nx^n \in P_n$$

$$T: P_n \rightarrow P_{n+1}: T(\mathbf{p}) = T(p(x)) = xp(x) = c_0x + c_1x^2 + \dots + c_nx^{n+1}$$

- **Example 5: (Functions that are not linear transformations)**

$$(a) f(x) = \sin x \quad \sin(x_1 + x_2) \neq \sin(x_1) + \sin(x_2)$$

$$(b) f(x) = x^2 \quad (x_1 + x_2)^2 \neq x_1^2 + x_2^2$$

$$(c) f(x) = x + 1$$

$$f(x_1 + x_2) = x_1 + x_2 + 1$$

$$f(x_1) + f(x_2) = (x_1 + 1) + (x_2 + 1) = x_1 + x_2 + 2$$

$$f(x_1 + x_2) \neq f(x_1) + f(x_2)$$

$$(d) T(\mathbf{v}) = \|\mathbf{v}\| \quad \|\mathbf{u} + \mathbf{v}\| \leq \|\mathbf{u}\| + \|\mathbf{v}\| \Rightarrow T(\mathbf{u} + \mathbf{v}) \neq T(\mathbf{u}) + T(\mathbf{v})$$

- **Zero transformation:**  $T: V \rightarrow W \quad T(\mathbf{v}) = \mathbf{0}, \quad \forall \mathbf{v} \in V$
- **Identity transformation:**  $T: V \rightarrow V \quad T(\mathbf{v}) = \mathbf{v}, \quad \forall \mathbf{v} \in V$
- **Theorem 1: (Properties of linear transformations)**

$T: V \rightarrow W, \quad \mathbf{u}, \mathbf{v} \in V$

$$(1) T(\mathbf{0}) = \mathbf{0} \quad (2) T(-\mathbf{v}) = -T(\mathbf{v}) \quad (3) T(\mathbf{u} - \mathbf{v}) = T(\mathbf{u}) - T(\mathbf{v})$$

(4) If  $\mathbf{v} = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \cdots + c_n\mathbf{v}_n$  then

$$T(\mathbf{v}) = T(c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \cdots + c_n\mathbf{v}_n) = c_1T(\mathbf{v}_1) + c_2T(\mathbf{v}_2) + \cdots + c_nT(\mathbf{v}_n)$$

- **Example 6: (Functions that are not linear transformations)**

Let  $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$  be a linear transformation such that

$$T(1, 0, 0) = (2, -1, 4), \quad T(0, 1, 0) = (1, 5, -2), \quad T(0, 0, 1) = (0, 3, 1)$$

Find  $T(2, 3, -2)$

$$(2, 3, -2) = 2(1, 0, 0) + 3(0, 1, 0) - 2(0, 0, 1)$$

$$T(2, 3, -2) = 2T(1, 0, 0) + 3T(0, 1, 0) - 2T(0, 0, 1)$$

$$= 2(2, -1, 4) + 3(1, 5, -2) - 2T(0, 3, 1) = (7, 7, 0)$$

■ **Example 7: (A linear transformation defined by a matrix)**

The function  $T: R^2 \rightarrow R^3$  is defined as  $T(\mathbf{v}) = A\mathbf{v} = \begin{bmatrix} 3 & 0 \\ 2 & 1 \\ -1 & -2 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$

(a) Find  $T(\mathbf{v})$ , where  $\mathbf{v} = (2, -1)$

(b) Show that  $T$  is a linear transformation from  $R^2$  into  $R^3$

$$(a) \mathbf{v} = (2, -1) \quad T(\mathbf{v}) = A\mathbf{v} = \begin{bmatrix} 3 & 0 \\ 2 & 1 \\ -1 & -2 \end{bmatrix} \begin{bmatrix} \overset{R^2 \text{ vector}}{\downarrow} 2 \\ -1 \end{bmatrix} = \begin{bmatrix} 6 \\ 3 \\ 0 \end{bmatrix} \Rightarrow T(2, -1) = (6, 3, 0)$$



$$(b) \quad T(\mathbf{u} + \mathbf{v}) = A(\mathbf{u} + \mathbf{v}) = A\mathbf{u} + A\mathbf{v} = T(\mathbf{u}) + T(\mathbf{v}) \quad (\text{vector addition})$$

$$T(c\mathbf{u}) = A(c\mathbf{u}) = c(A\mathbf{u}) = cT(\mathbf{u}) \quad (\text{scalar multiplication})$$

- **Theorem 2: (The linear transformation given by a matrix)**

Let  $A$  be an  $m \times n$  matrix. The function  $T$  defined by  $T(\mathbf{v}) = A\mathbf{v}$  is a linear transformation from  $R^n$  into  $R^m$ .

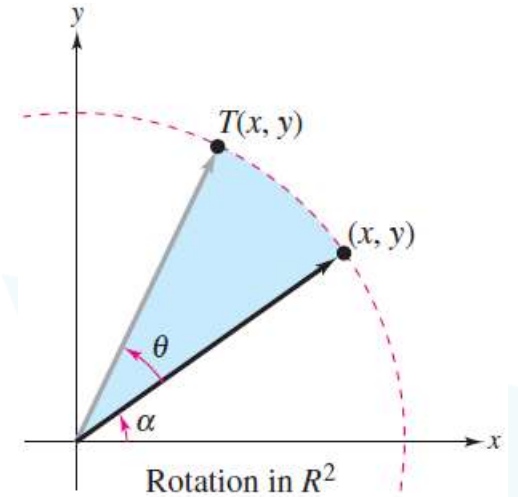
- **Example 8: (Rotation in the plane)**

Show that the LT  $T: R^2 \rightarrow R^2$  given by the matrix  $A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$

has the property that it rotates every vector in  $R^2$  counterclockwise about the origin through the angle  $\theta$ .

$$\mathbf{v} = (x, y) = (r \cos \alpha, r \sin \alpha) \quad (\text{polar coordinates})$$

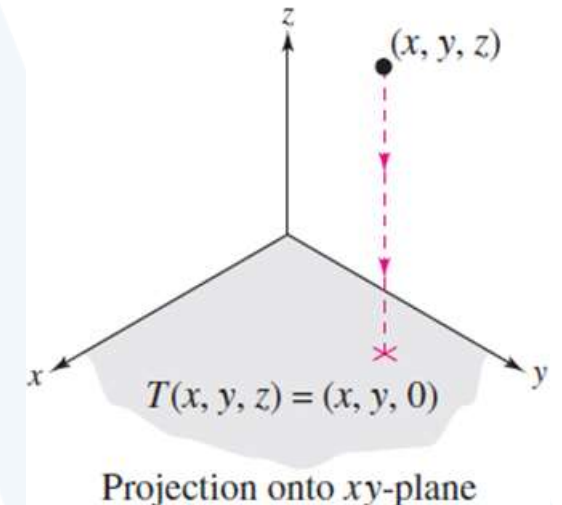
$$\begin{aligned}
 T(\mathbf{v}) = A\mathbf{v} &= \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} r \cos\alpha \\ r \sin\alpha \end{bmatrix} \\
 &= \begin{bmatrix} r \cos\theta \cos\alpha - r \sin\theta \sin\alpha \\ r \sin\theta \cos\alpha + r \cos\theta \sin\alpha \end{bmatrix} \\
 &= \begin{bmatrix} r \cos(\theta + \alpha) \\ r \sin(\theta + \alpha) \end{bmatrix}
 \end{aligned}$$



Thus,  $T(\mathbf{v})$  is the vector that results from **rotating** the vector  $\mathbf{v}$  **counterclockwise** through the angle  $\theta$ .

- **Example 9: (A projection in  $R^3$ )**

The LTT:  $R^3 \rightarrow R^3$  is given by the matrix  $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$  is called a projection in  $R^3$ .



## 2. The Kernel and Range of a Linear Transformation

- **Definition:** Let  $T: V \rightarrow W$  be a Linear transformation. Then the set of all vectors  $v$  in  $V$  that satisfy  $T(v) = \mathbf{0}$  is called the **kernel** of  $T$  and is denoted by  $\ker(T)$ .

$$\ker(T) = \{v \mid T(v) = \mathbf{0}, \forall v \in V\}$$

- **Example 10: (The kernel of the zero and identity transformations)**

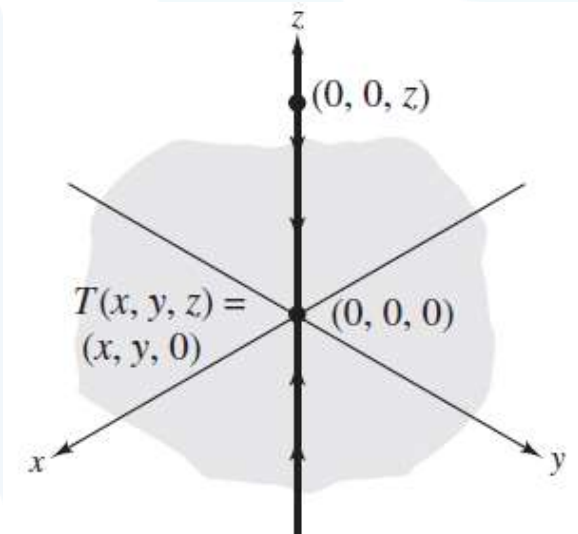
(a)  $T(v) = \mathbf{0}$  (the zero transformation)       $\ker(T) = V$

(b)  $T(v) = v$  (the identity transformation)       $\ker(T) = \{\mathbf{0}\}$

- **Example 11: (Finding the kernel of a LT)**

$$T(v) = (x, y, 0) \quad T: R^3 \rightarrow R^3$$

$$\ker(T) = \{(0, 0, z) \mid z \text{ is a real number}\}$$



- Example 12: (Finding the kernel of a linear transformation)

$$T(\mathbf{x}) = A\mathbf{x} = \begin{bmatrix} 1 & -1 & -2 \\ -1 & 2 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \quad (T: R^3 \rightarrow R^2)$$

$$\ker(T) = \{(x_1, x_2, x_3) \mid T(x_1, x_2, x_3) = (0, 0), \mathbf{x} = (x_1, x_2, x_3) \in R^3\}$$

$$T(x_1, x_2, x_3) = (0, 0) \Rightarrow \begin{bmatrix} 1 & -1 & -2 \\ -1 & 2 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & -1 & -2 & 0 \\ -1 & 2 & 3 & 0 \end{bmatrix} \xrightarrow{\text{Gauss-J. Elimination}} \begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 1 & 0 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} t \\ -t \\ t \end{bmatrix} = t \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

$$\Rightarrow \ker(T) = \{t(1, -1, 1) \mid t \text{ is a real number}\} = \text{span}\{(1, -1, 1)\}$$

- Theorem 3: (The kernel is a subspace of  $V$ )**

The kernel of a linear transformation  $T: V \rightarrow W$  is a subspace of the domain  $V$ .

- Definition:** Let  $T: V \rightarrow W$  be a Linear transformation. Then the set of all vectors  $w$  in  $W$  that are images of vectors in  $V$  is called the **range** of  $T$  and is denoted by  $\text{range}(T)$  or  $R(T)$ .  

$$\text{range}(T) = \{T(v) \mid \forall v \in V\}$$

- Theorem 4: (The range is a subspace of  $W$ )**

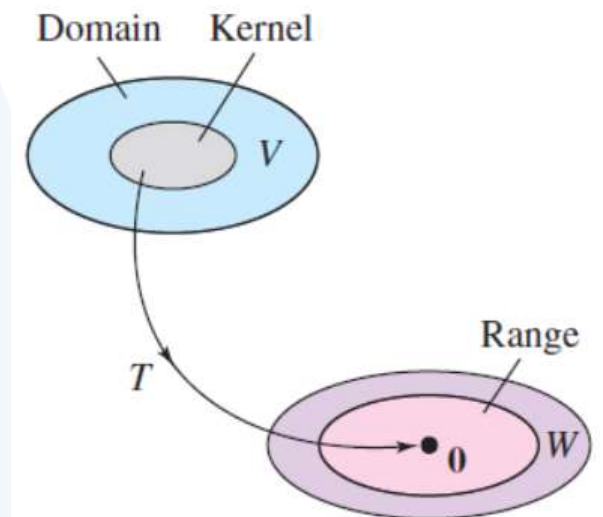
The range of a LT  $T: V \rightarrow W$  is a subspace of the  $W$ .

- Rank of a linear transformation  $T: V \rightarrow W$ :**

$\text{rank}(T)$  = the dimension of the range of  $T$

- Nullity of a linear transformation  $T: V \rightarrow W$ :**

$\text{nullity}(T)$  = the dimension of the kernel of  $T$



- **Note:** Let  $T: R^n \rightarrow R^m$  be the LT given by  $T(\mathbf{x}) = A\mathbf{x}$ . Then  
 $\text{rank}(T) = \text{rank}(A),$                        $\text{nullity}(T) = \text{nullity}(A)$

- **Theorem 5: (Sum of rank and nullity)**

Let  $T: V \rightarrow W$  be a LT from an  $n$ -dimensional vector space  $V$  into a vector space  $W$ . Then:

$$\text{rank}(T) + \text{nullity}(T) = n$$

$$\text{dim}(\text{range of } T) + \text{dim}(\text{kernel of } T) = \text{dim}(\text{domain of } T)$$

- **Example 13: (Finding rank and nullity of a linear transformation)**

Find the rank and nullity of the LT  $T: R^3 \rightarrow R^3$  defined by

$$\text{rank}(T) = \text{rank}(A) = 2$$

$$\text{nullity}(T) = \text{dim}(\text{domain of } T) - \text{rank}(T) = 3 - 2 = 1$$

$$A = \begin{bmatrix} 1 & 0 & -2 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

- **Example 14: (Finding rank and nullity of a linear transformation)**

Let  $T: R^5 \rightarrow R^7$  be a linear transformation

(a) Find the dimension of the kernel of  $T$  if the dimension of the range is 2

(b) Find the rank of  $T$  if the nullity of  $T$  is 4

(c) Find the rank of  $T$  if  $\ker(T) = \{\mathbf{0}\}$

(a)  $\dim(\text{domain of } T) = 5 \Rightarrow \dim(\ker \text{ of } T) = n - \dim(\text{range of } T) = 5 - 2 = 3$

(b)  $\text{rank}(T) = n - \text{nullity}(T) = 5 - 4 = 1$       (c)  $\text{rank}(T) = n - \text{nullity}(T) = 5 - 0 = 5$

### 3. Compositions and Inverse Transformations

- **Definition:** A function  $T: V \rightarrow W$  is **one-to-one** when the preimage of every  $w$  in the range consists of a single vector.

$T$  is one-to-one if and only if, for all  $u$  and  $v$  in  $V$ ,  $T(u) = T(v)$  implies  $u = v$ .

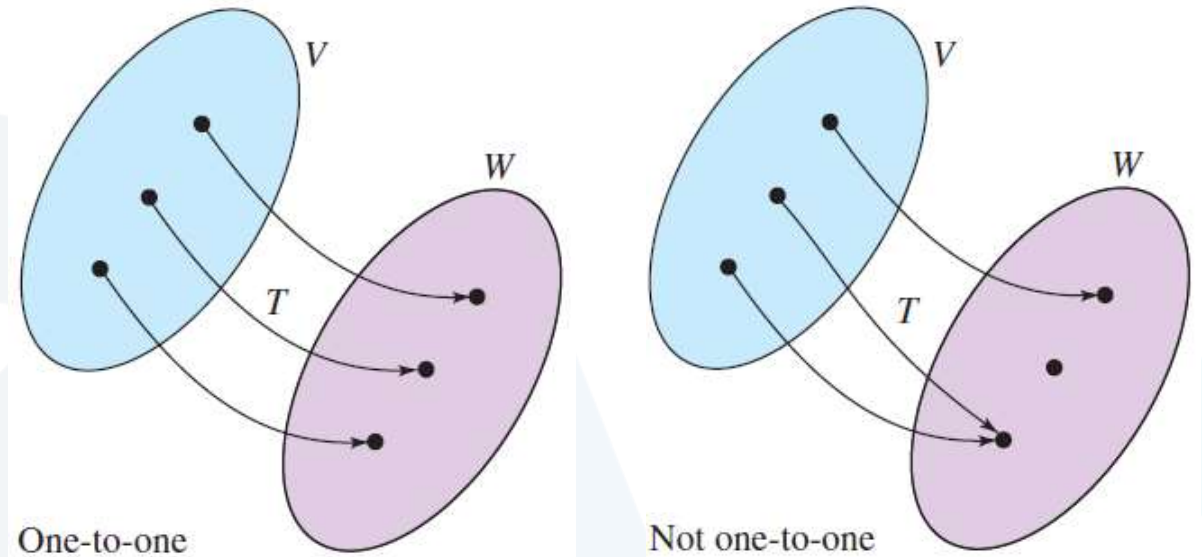
- **Definition:** A function  $T: V \rightarrow W$  is **onto** when every element in  $W$  has a preimage in  $V$ . ( $T$  is onto  $W$  when  $W$  is equal to the range of  $T$ ).

- **Theorem 6: (One-to-one LT)**

Let  $T: V \rightarrow W$  be a LT. Then  $T$  is one-to-one iff  $\ker(T) = \{\mathbf{0}\}$ .

- **Example 15: (One-to-one and not one-to-one linear transformation)**

- The linear transformation  $T: M_{3 \times 2} \rightarrow M_{2 \times 3}$  given by  $T(A) = A^T$  is one-to-one because its kernel consists of only the  $m \times n$  zero matrix.
- The zero transformation  $T: R^3 \rightarrow R^3$  is not one-to-one because its kernel is all of  $R^3$ .





- **Example 16: (One-to-one and onto linear transformation)**

The LT  $T: P_3 \rightarrow R^4$  given by  $T(a + bx + cx^2 + dx^3) = (a, b, c, d)$ .

- **Example 17: (One-to-one and not onto linear transformation)**

$$T: P_n \rightarrow P_{n+1}: T(p) = T(p(x)) = xp(x)$$

- **Theorem 7: (Onto linear transformation)**

Let  $T: V \rightarrow W$  be a linear transformation, where  $W$  is finite dimensional. Then  $T$  is onto iff the rank of  $T$  is equal to the dimension of  $W$ .

- **Theorem 8: (One-to-one and onto linear transformation)**

Let  $T: V \rightarrow W$  be a linear transformation, with vector space  $V$  and  $W$  both of dimension  $n$ . Then  $T$  is one-to-one iff it is onto.

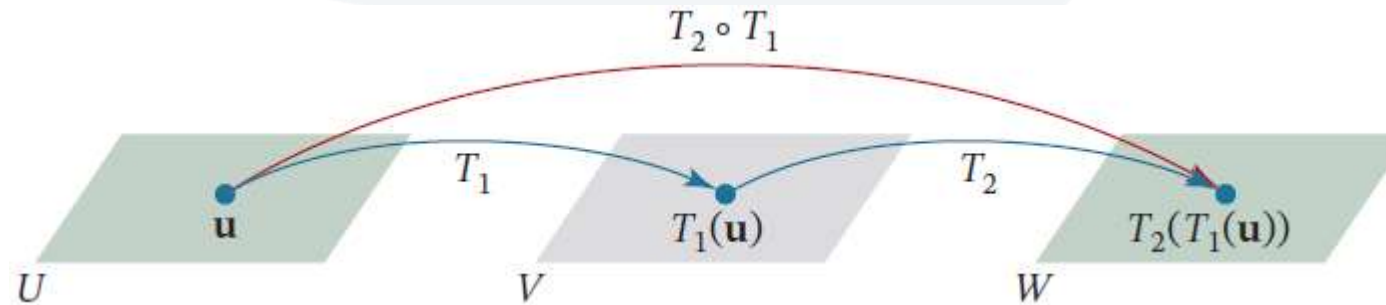
■ **Example 18: (One-to-one and not one-to-one linear transformation)**

Let  $T: R^n \rightarrow R^m$  be a LT given by  $T(\mathbf{x}) = A\mathbf{x}$ . Find the nullity and rank of  $T$  to determine whether  $T$  is one-to-one, onto, or neither.

$$(a) A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}, (b) A = \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}, (c) A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 1 & -1 \end{bmatrix}, (d) A = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

$T: R^n \rightarrow R^m$	$\dim(\text{domain of } T)$	$\text{rank}(T)$	$\text{nullity}(T)$	one-to-one	onto
(a) $T: R^3 \rightarrow R^3$	3	3	0	Yes	Yes
(b) $T: R^2 \rightarrow R^3$	2	2	0	Yes	No
(c) $T: R^3 \rightarrow R^2$	3	2	1	No	Yes
(d) $T: R^3 \rightarrow R^3$	3	2	1	No	No

**Definition:** If  $T_1: U \rightarrow V$  and  $T_2: V \rightarrow W$  are linear transformations, then the composition of  $T_2$  with  $T_1$ , denoted by  $T_2 \circ T_1$  is the function defined by the formula  $(T_2 \circ T_1)(u) = T_2(T_1(u))$ , where  $u$  is a vector in  $U$ .



- **Theorem 9: (Composition of linear transformations)**

If  $T_1: U \rightarrow V$  and  $T_2: V \rightarrow W$  are linear transformations, then  $(T_2 \circ T_1): U \rightarrow W$  is also a linear transformations.

- **Example 19: (Composition of linear transformations)**

Let  $T_1$  and  $T_2$  be linear transformations from  $R^3$  into  $R^3$  such that:

$$T_1(x, y, z) = (2x + y, 0, x + z), \quad T_2(x, y, z) = (x - y, z, y)$$

Find the compositions  $T = T_2 \circ T_1$  and  $T' = T_1 \circ T_2$

$$(T_2 \circ T_1)(x, y, z) = T_2(T_1(x, y, z)) = T_2(2x + y, 0, x + z) = (2x + y, x + z, 0)$$

$$(T_1 \circ T_2)(x, y, z) = T_1(T_2(x, y, z)) = T_1(x - y, z, y) = (2x - 2y + z, 0, x)$$

- **Note:**  $T_2 \circ T_1 \neq T_1 \circ T_2$
- **Composition with the Identity Operator**

If  $T: V \rightarrow V$  is any linear operator, and if  $I: V \rightarrow V$  is the identity, then for all vectors  $v$  in  $V$ , we have

$$(T \circ I)(v) = T(I(v)) = T(v)$$

$$(I \circ T)(v) = I(T(v)) = T(v)$$

$$T \circ I = T \text{ and } I \circ T = T$$

- Note:** Let  $T_1: R^n \rightarrow R^m$  and  $T_2: R^m \rightarrow R^p$  be LT where  $T_1(\mathbf{u}) = A_1\mathbf{u}$  and  $T_2(\mathbf{v}) = A_2\mathbf{v}$ , then
  - The composition  $T: R^n \rightarrow R^p$ , defined by  $T(\mathbf{v}) = T_2(T_1(\mathbf{v}))$ , is a LT.
  - The matrix  $A$  for  $T$  is given the matrix product  $A = A_2A_1$ , where  $T(\mathbf{u}) = A\mathbf{u}$
- Example 20: (Composition of linear transformations)**

Let  $T_1$  and  $T_2$  be linear transformations from  $R^3$  into  $R^3$  such that:

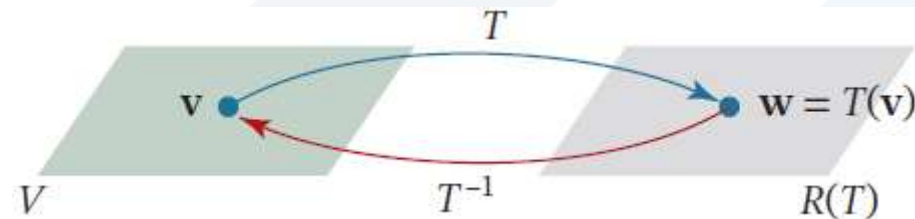
$$T_1(x, y, z) = (2x + y, 0, x + z), \quad T_2(x, y, z) = (x - y, z, y)$$

Find the composition  $T = T_2 \circ T_1$

$$A_1 = \begin{bmatrix} 2 & 1 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{bmatrix}, \quad A_2 = \begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \Rightarrow A = A_2A_1 = \begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

$$T(x, y, z) = A \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 2 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = (2x + y, x + z, 0)$$

**Definition:** If  $T: V \rightarrow W$  is a linear transformations, then  $T$  is **invertible** if there is a transformation  $T^{-1}$  such that:  $T^{-1} \circ T = I_V$  and  $T \circ T^{-1} = I_W$ . We call  $T^{-1}$  the **inverse** of  $T$ .

$$T^{-1}(T(u)) = u, \quad \forall u \in U \qquad T(T^{-1}(w)) = w, \quad \forall w \in R(T)$$


■ **Notes:**

- (1) The inverse transformation  $T^{-1}: R(T) \rightarrow V$  exists iff  $T$  is **one-to-one**.
- (2) If  $T: V \rightarrow W$  is a **linear transformations**, then  $T^{-1}: R(T) \rightarrow V$  is also a **LT**.

- **Example 21: (An Inverse Transformation)**

$$T: P_n \rightarrow P_{n+1}: T(p) = T(p(x)) = xp(x) = c_0x + c_1x^2 + \dots + c_nx^{n+1}$$

is a one-to-one LT  $\Rightarrow T^{-1}(c_0x + c_1x^2 + \dots + c_nx^{n+1}) = c_0 + c_1x + \dots + c_nx^n$

- **Note:** Consider  $T: R^n \rightarrow R^n$  where  $T(u) = Au$

(1)  $T$  is one-to-one if and only if  $A$  is invertible.

(2)  $T^{-1}$  exists if and only if  $A$  is invertible.

The inverse transformation is the matrix transformation given by  $A^{-1}$ .

- **Example 22: (Finding the inverse of a linear transformation)**

The linear transformations  $T: R^3 \rightarrow R^3$  defined by:

$$T(x_1, x_2, x_3) = (2x_1 + 3x_2 + x_3, 3x_1 + 3x_2 + x_3, 2x_1 + 4x_2 + x_3)$$

Show that  $T$  is invertible, and find its inverse.

## 4. Geometry of Matrix Operators

### ■ Example 23: (Transformation of the Unit Square)

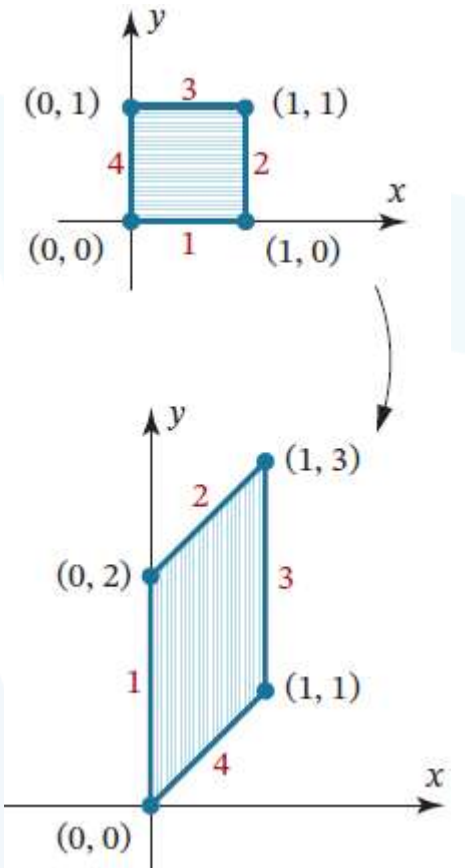
Sketch the image of the unit square under multiplication by the invertible matrix:

$$A = \begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \quad \begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad \begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

The image of the unit square is a **parallelogram** with vertices  $(0, 0)$ ,  $(0, 2)$ ,  $(1, 1)$ , and  $(1, 3)$ .

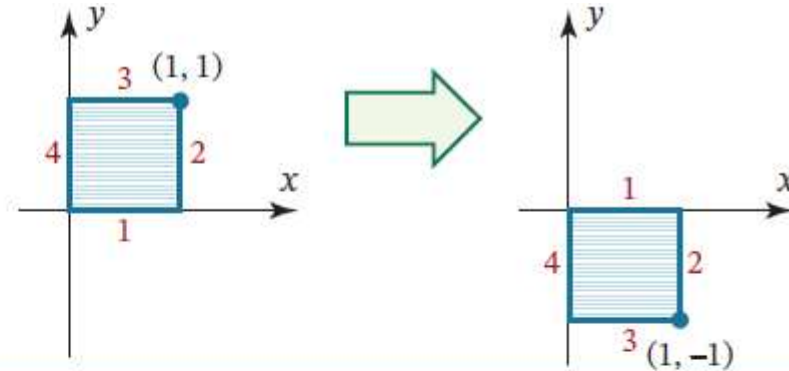




# Reflections, Rotations, and Projections

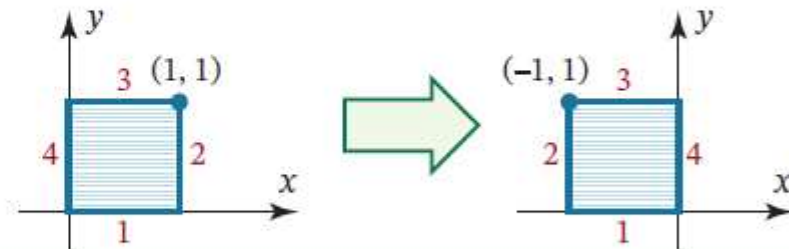
Reflection about  
the  $x$ -axis

$$T(x, y) = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} (x, -y)$$



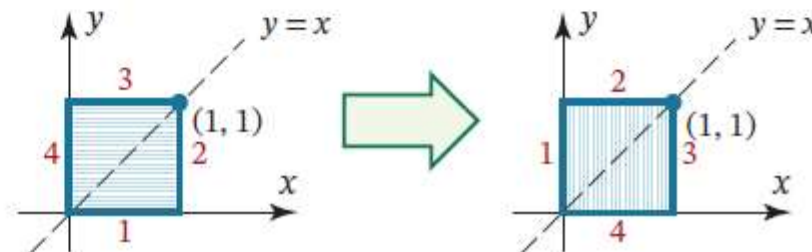
Reflection about  
the  $y$ -axis

$$T(x, y) = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} (-x, y)$$



Reflection about  
the line  $y = x$

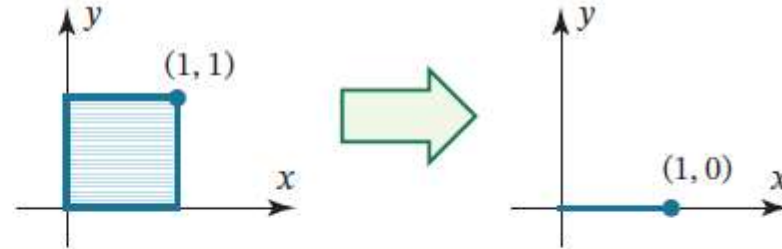
$$T(x, y) = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} (y, x)$$



Orthogonal  
projection  
onto the  $x$ -axis

$$\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

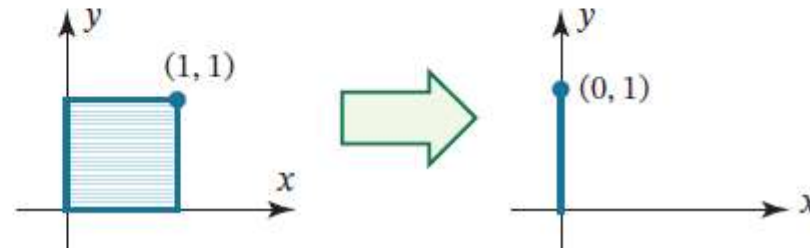
$$T(x, y) = (x, 0)$$



Orthogonal  
projection  
onto the  $y$ -axis

$$\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

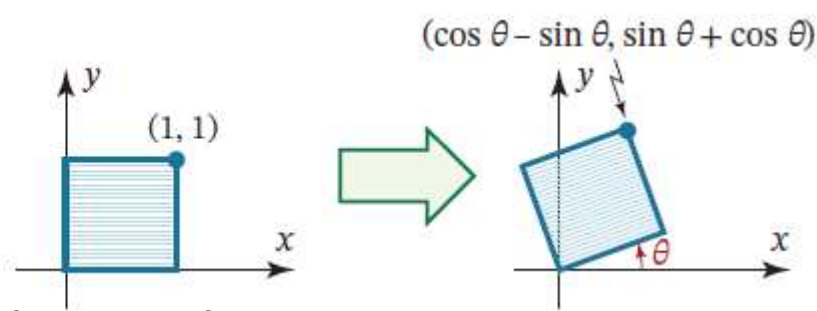
$$T(x, y) = (0, y)$$



Rotation about the  
origin through a  
positive angle  $\theta$

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

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

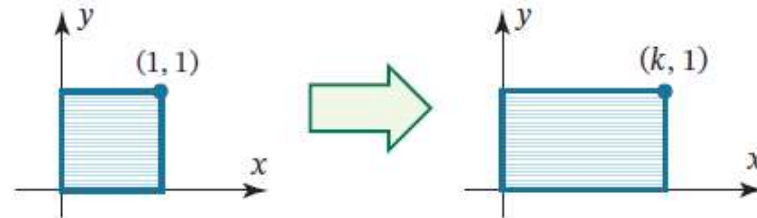


## Expansions and Compressions

Expansion in the  
 $x$ -direction with  
factor  $k$

$$\begin{bmatrix} k & 0 \\ 0 & 1 \end{bmatrix}$$

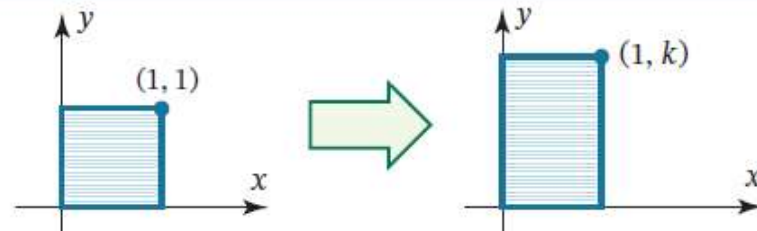
$$(k > 1) \quad T(x, y) = (kx, y)$$



Expansion in the  
 $y$ -direction with  
factor  $k$

$$\begin{bmatrix} 1 & 0 \\ 0 & k \end{bmatrix}$$

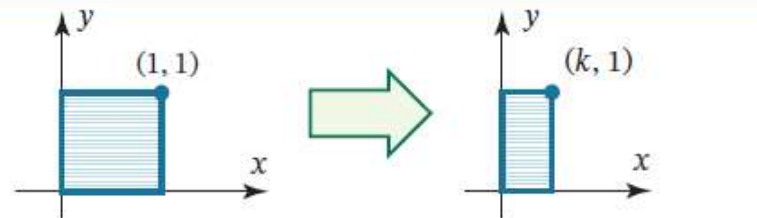
$$(k > 1) \quad T(x, y) = (x, ky)$$



Compression in the  
 $x$ -direction with  
factor  $k$

$$\begin{bmatrix} k & 0 \\ 0 & 1 \end{bmatrix}$$

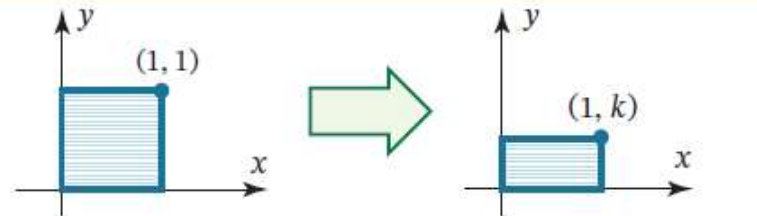
$$(0 < k < 1)$$



Compression in the  
 $y$ -direction with  
factor  $k$

$$\begin{bmatrix} 1 & 0 \\ 0 & k \end{bmatrix}$$

$$(0 < k < 1)$$

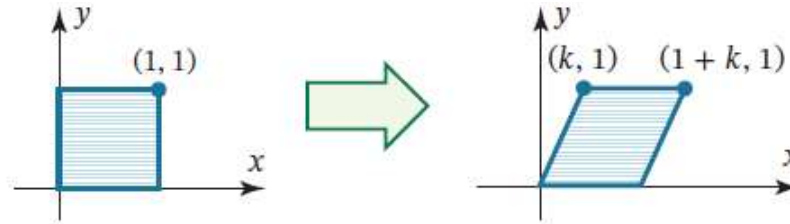


# Shears

Shear in the  
positive  $x$ -direction  
by a factor  $k$

$$\begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}$$

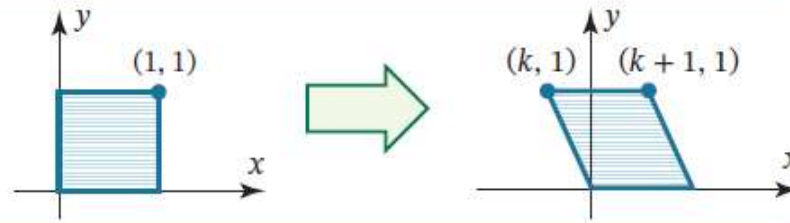
$$(k > 0) \quad T(x, y) = (x + ky, y)$$



Shear in the  
negative  $x$ -direction  
by a factor  $k$

$$\begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}$$

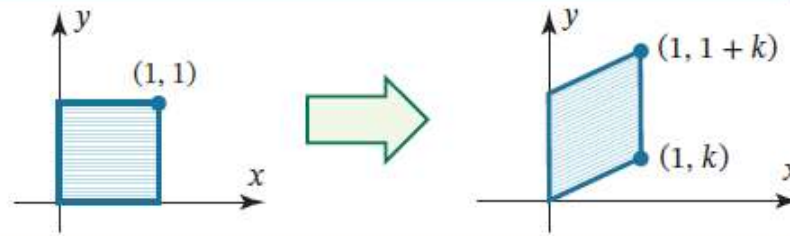
$$(k < 0)$$



Shear in the  
positive  $y$ -direction  
by a factor  $k$

$$\begin{bmatrix} 1 & 0 \\ k & 1 \end{bmatrix}$$

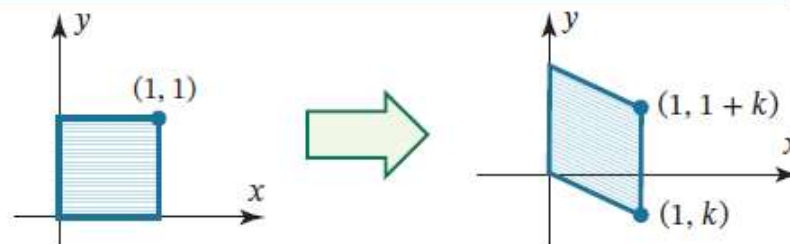
$$(k > 0) \quad T(x, y) = (x, kx + y)$$



Shear in the  
negative  $y$ -direction  
by a factor  $k$

$$\begin{bmatrix} 1 & 0 \\ k & 1 \end{bmatrix}$$

$$(k < 0)$$



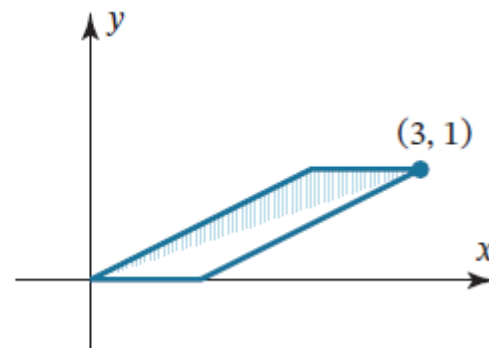
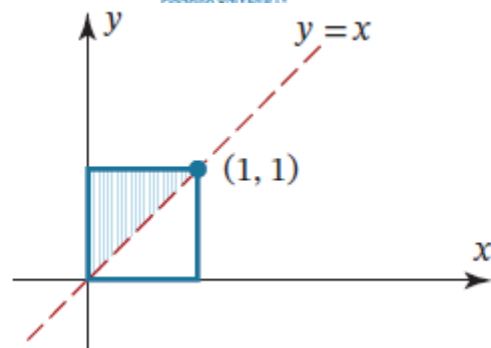
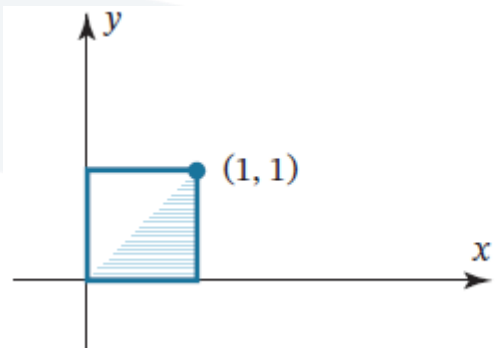
■ **Example 24: (Transformation of the Unit Square)**

(a) Find the standard matrix for the operator on  $R^2$  that first shears by a factor of 2 in the  $x$ -direction and then reflects the result about the line  $y = x$ . Sketch the image of the unit square under this operator.

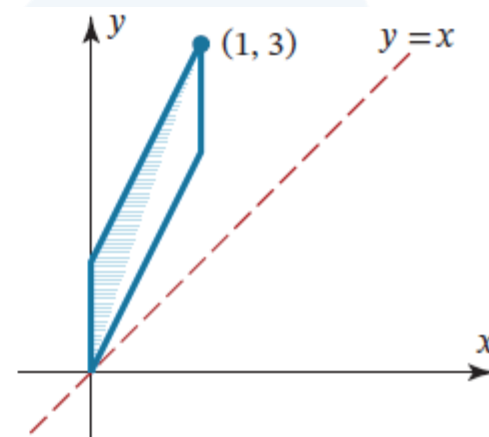
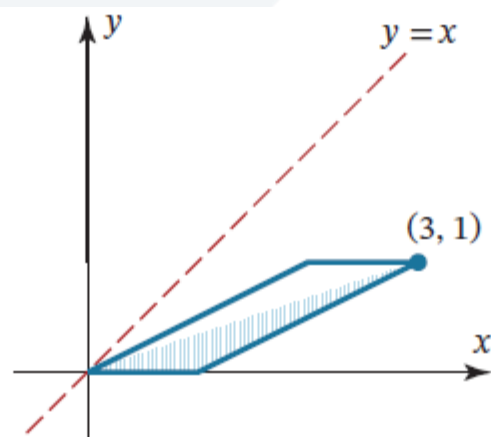
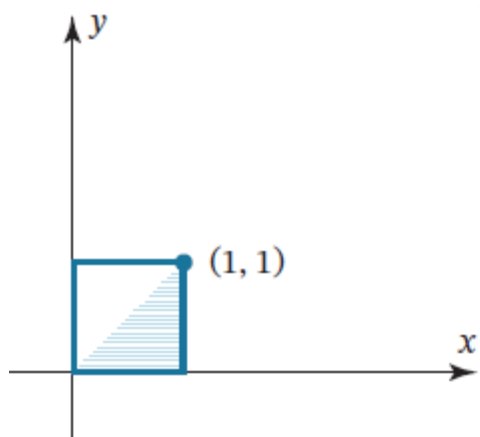
(b) Find the standard matrix for the operator on  $R^2$  that first reflects about  $y = x$  and then shears by a factor of 2 in the  $x$ -direction. Sketch the image of the unit square under this operator. Conclude.

(a) The matrix for the shear is  $A_1 = \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix}$  and for the reflection is  $A_2 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

$$A_2 A_1 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 2 \end{bmatrix}$$



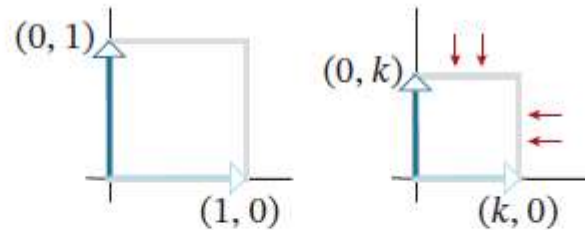
$$(b) A_1 A_2 = \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 1 & 0 \end{bmatrix}$$



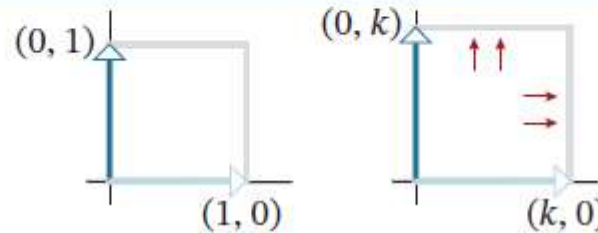
$$A_1 A_2 \neq A_2 A_1$$

## Dilations and Contractions

Contraction with  
factor  $k$  in  $\mathbb{R}^2$   
( $0 \leq k < 1$ )



Dilation with  
factor  $k$  in  $\mathbb{R}^2$   
( $k > 1$ )



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

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

$$A = \begin{bmatrix} k_1 & 0 \\ 0 & k_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & k_2 \end{bmatrix} \begin{bmatrix} k_1 & 0 \\ 0 & 1 \end{bmatrix}$$

- Note:** The multiplication by  $A$  causes a **compression** or **expansion** of the unit square by a factor of  $k_1$  in the  **$x$ -direction** followed by an **expansion** or **compression** of the unit square by a factor of  $k_2$  in the  **$y$ -direction**.



- **Reflection About the Origin:**

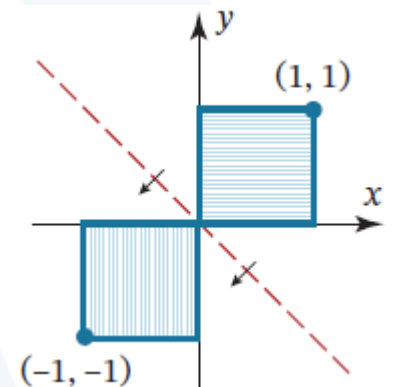
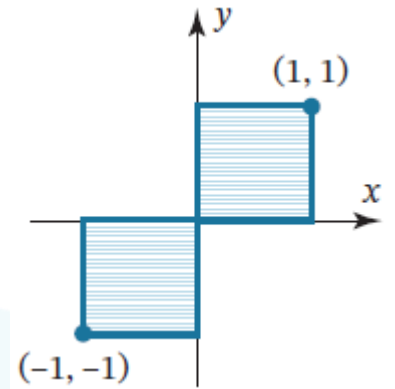
Multiplication by the matrix  $A = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}$  has the geometric effect of **reflecting** the unit square about the **origin**.

$$A = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

The same result can be obtained by first **reflecting** the unit square about the ***x*-axis** and then **reflecting** that result about the ***y*-axis**.

- **Reflection About the Line  $y = -x$**

$$A = \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$$





- **Theorem 10: (Elementary matrix transformations)**

If  $E$  is an elementary matrix, then  $T_E: R^2 \rightarrow R^2$  is one of the following:

- (a) A **shear** along a coordinate axis.
- (b) A **reflection** about  $y = x$ .
- (c) A **compression** along a coordinate axis.
- (d) An **expansion** along a coordinate axis.
- (e) A **reflection** about a coordinate axis.
- (f) A **compression** or **expansion** along a coordinate axis followed by a reflection about a coordinate axis.

- **Theorem 11: (Invertible matrix transformations)**

If  $T_A: R^2 \rightarrow R^2$  is multiplication by an invertible matrix  $A$ , then the geometric effect of  $T_A$  is the same as an appropriate succession of **shears**, **compressions**, **expansions**, and **reflections**.

- **Example 25: (Decomposing a Matrix Operator)**

In **Example 23** we illustrated the effect on the unit square of multiplication by:

$$A = \begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix}$$

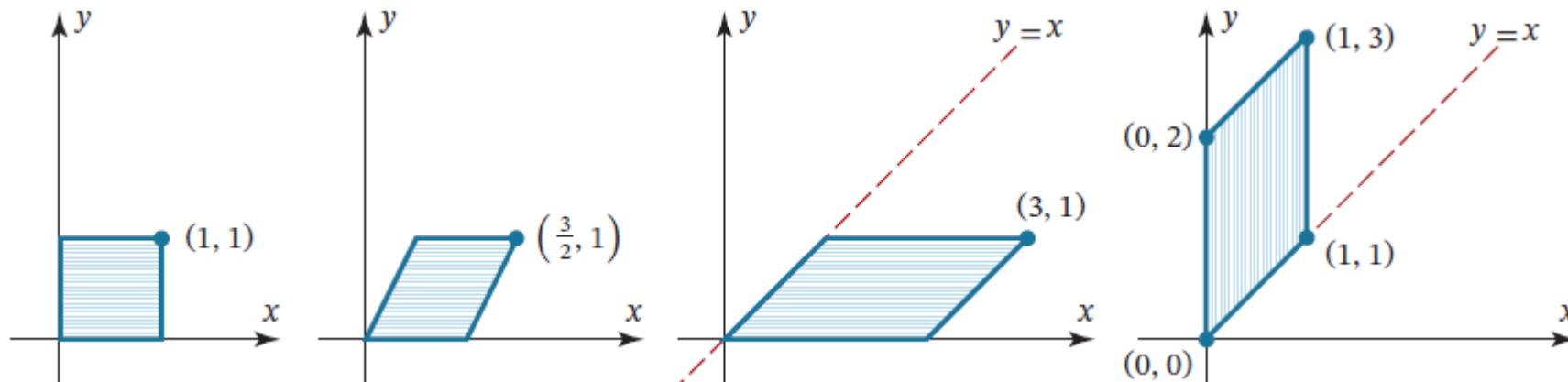
Express this matrix as a product of elementary matrices, and then describe the effect of multiplication by the matrix  $A$  in terms of shears, compressions, expansions, and reflections.

$$A = \begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix} \xrightarrow{r_{12}} \begin{bmatrix} 2 & 1 \\ 0 & 1 \end{bmatrix} \xrightarrow{r_1^{(1/2)}} \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & 1 \end{bmatrix} \xrightarrow{r_{21}^{(-1/2)}} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$$

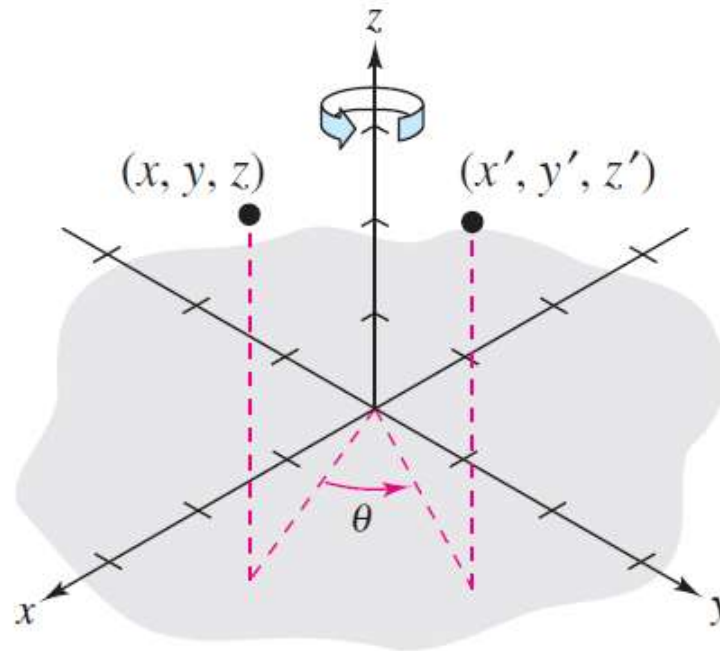
$$A = \begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix} = E_1^{-1} E_2^{-1} E_3^{-1} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & 1 \end{bmatrix}$$

Reading from right to left we can now see that the geometric effect of multiplying by  $A$  is equivalent to successively:

1. **shearing** by a factor of  $\frac{1}{2}$  in the  $x$ -direction;
2. **expanding** by a factor of 2 in the  $x$ -direction;
3. **reflecting** about the line  $y = x$ .



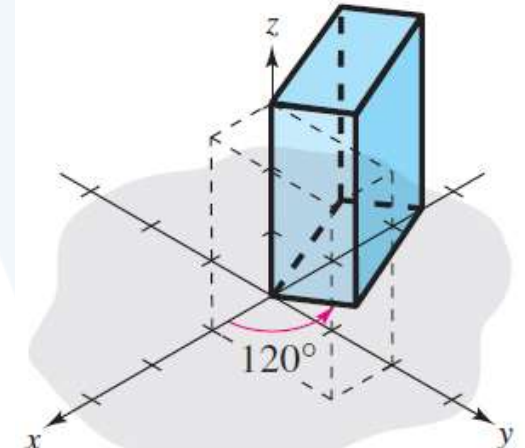
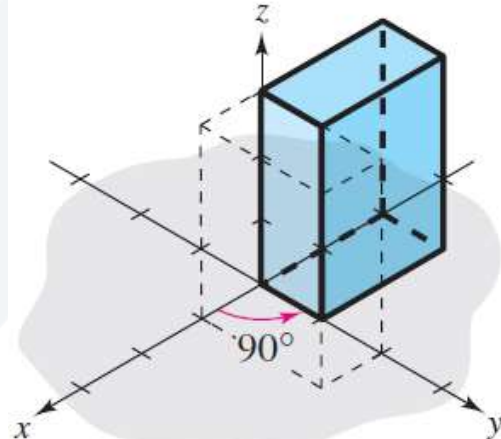
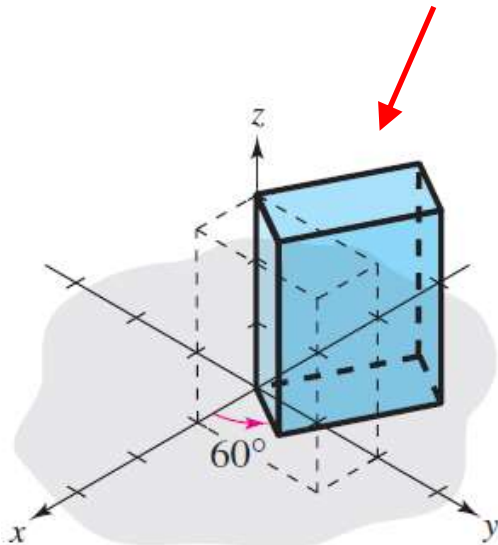
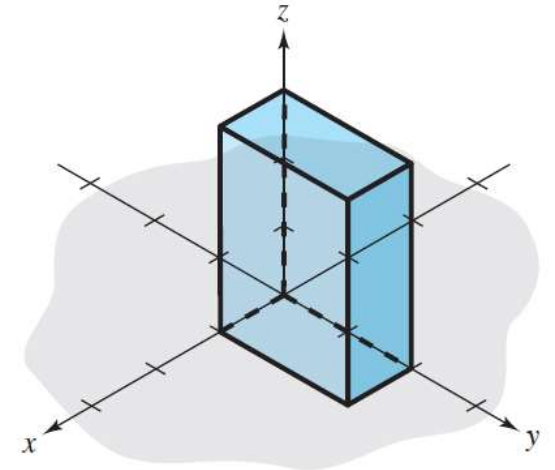
- Rotation In  $R^3$



$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} x \cos \theta - y \sin \theta \\ x \sin \theta + y \cos \theta \\ z \end{bmatrix}$$

## Rotation about the $z$ -axis

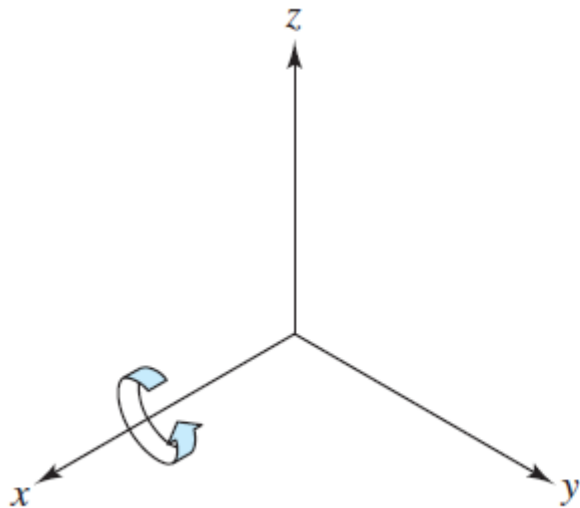
$$A = \begin{bmatrix} \cos 60^\circ & -\sin 60^\circ & 0 \\ \sin 60^\circ & \cos 60^\circ & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1/2 & -\sqrt{3}/2 & 0 \\ \sqrt{3}/2 & 1/2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$





### Rotation about the $x$ -axis

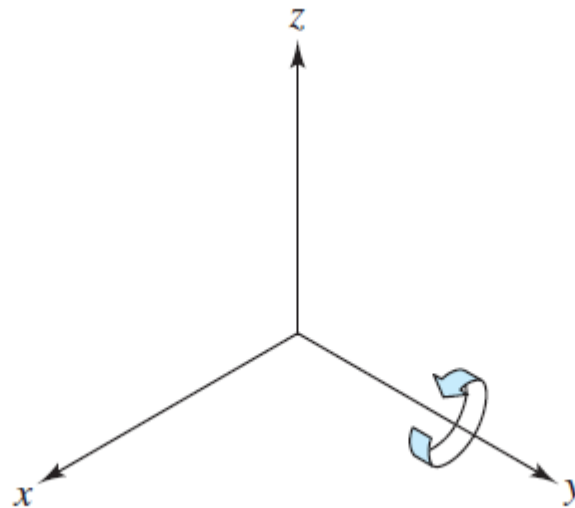
$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \theta & -\sin \theta \\ 0 & \sin \theta & \cos \theta \end{bmatrix}$$



Rotation about  $x$ -axis

### Rotation about the $y$ -axis

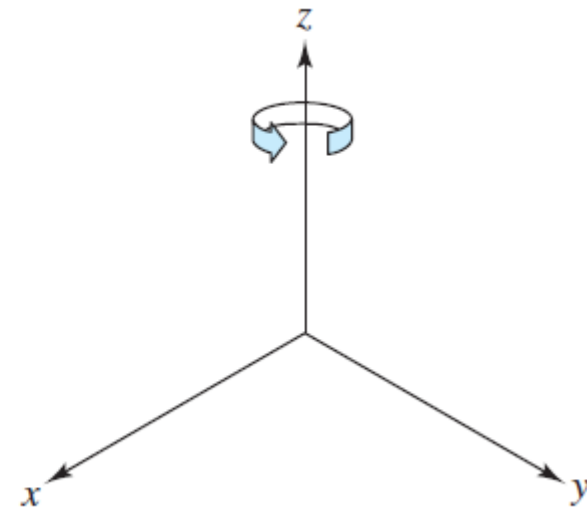
$$\begin{bmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{bmatrix}$$



Rotation about  $y$ -axis

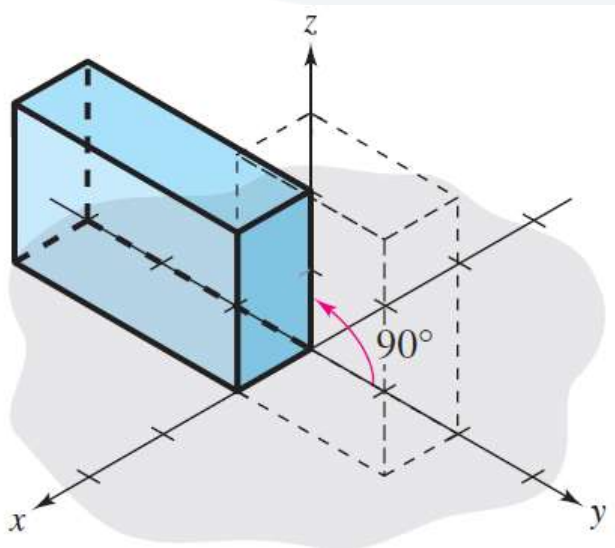
### Rotation about the $z$ -axis

$$\begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$$



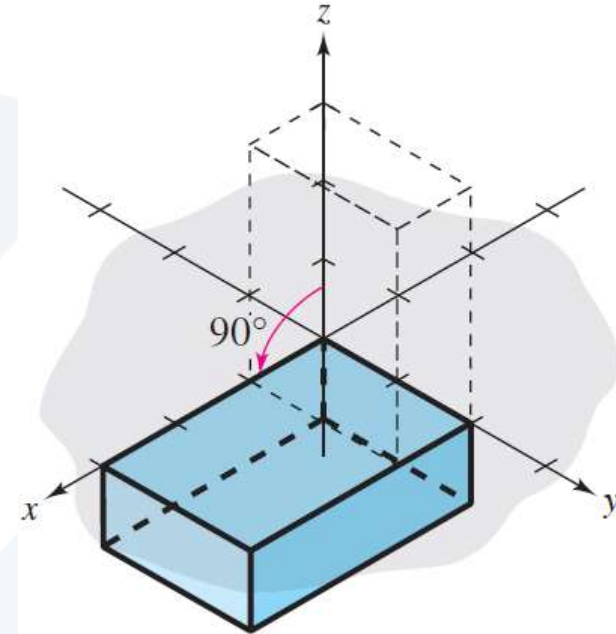
Rotation about  $z$ -axis

## Rotation of $90^\circ$ about the $x$ -axis



$$A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 0 \end{bmatrix}$$

## Rotation of $90^\circ$ about the $y$ -axis



$$A = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ -1 & 0 & 0 \end{bmatrix}$$