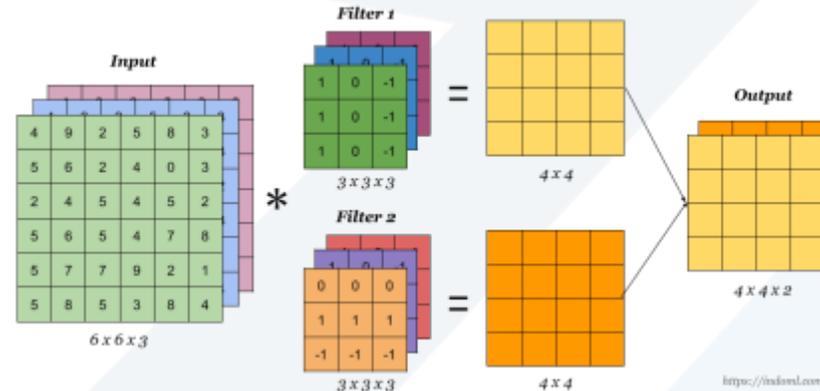


CECC102, CECC122 & CEDC102 : Linear Algebra (and Matrix Theory)

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

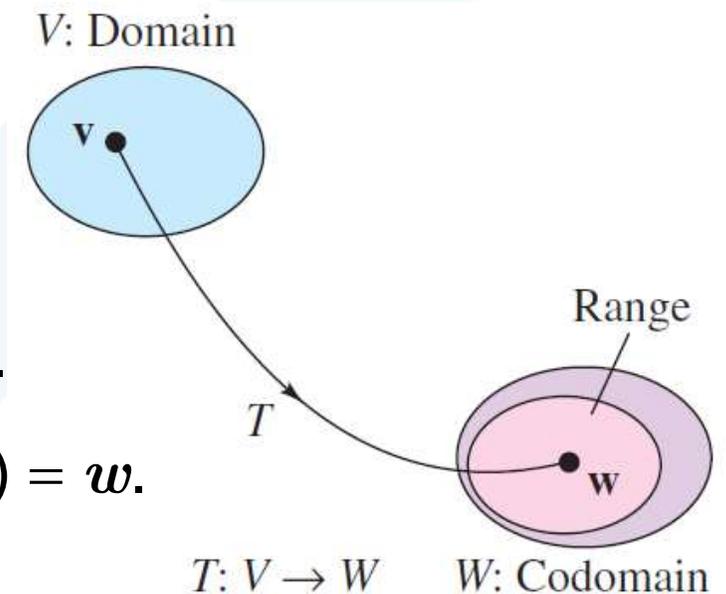
Images And Preimages of Functions:

- 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)**

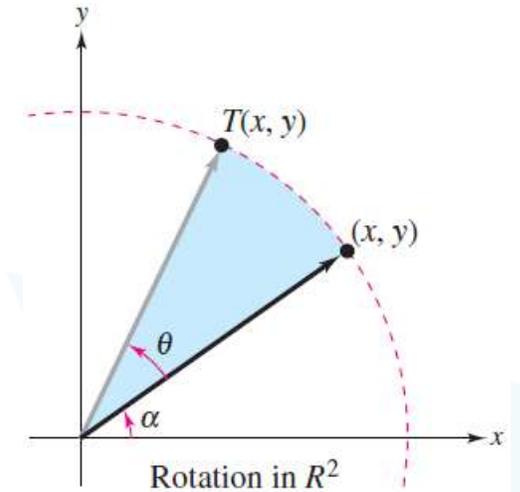
Show that the LT $T: R^2 \rightarrow R^2$ given by the matrix $A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$

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

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



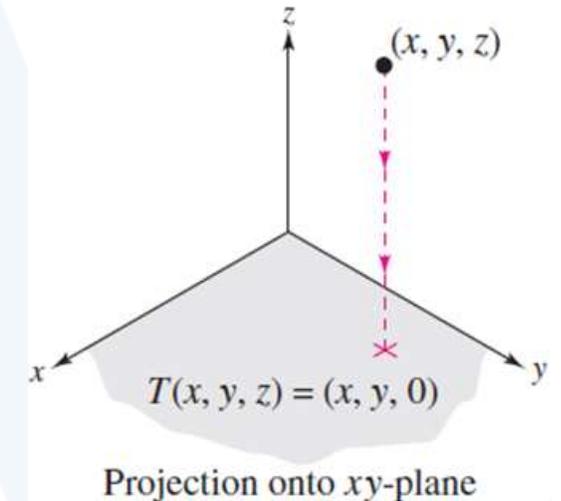
$$\begin{aligned}
 T(\mathbf{v}) = A\mathbf{v} &= \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} r \cos\alpha \\ r \sin\alpha \end{bmatrix}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



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

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

The LT $T: R^3 \rightarrow R^3$ 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)**

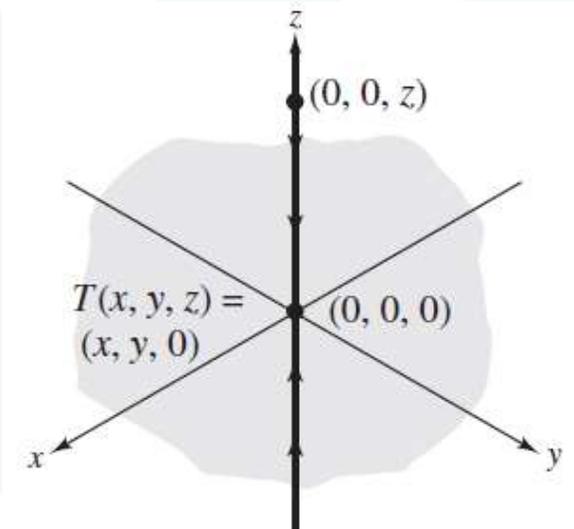
(a) $T(v) = \mathbf{0}$ (the zero transformation) $\ker(T) = V$

(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)**

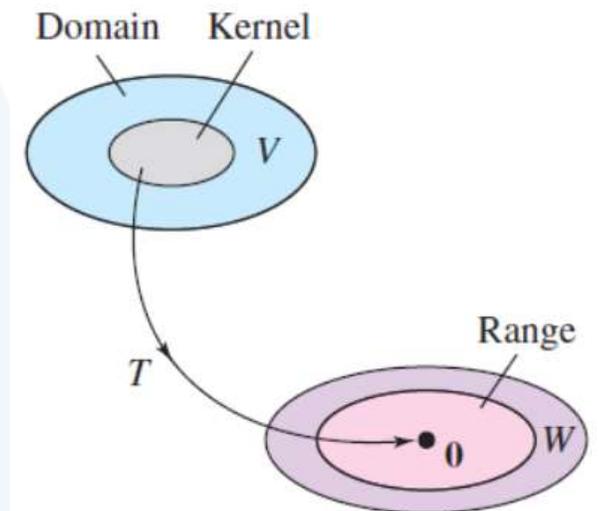
The range of a LT $T: V \rightarrow W$ is a subspace of the W .

- 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:

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

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

- **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

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

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

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



■ **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 $u = v$.



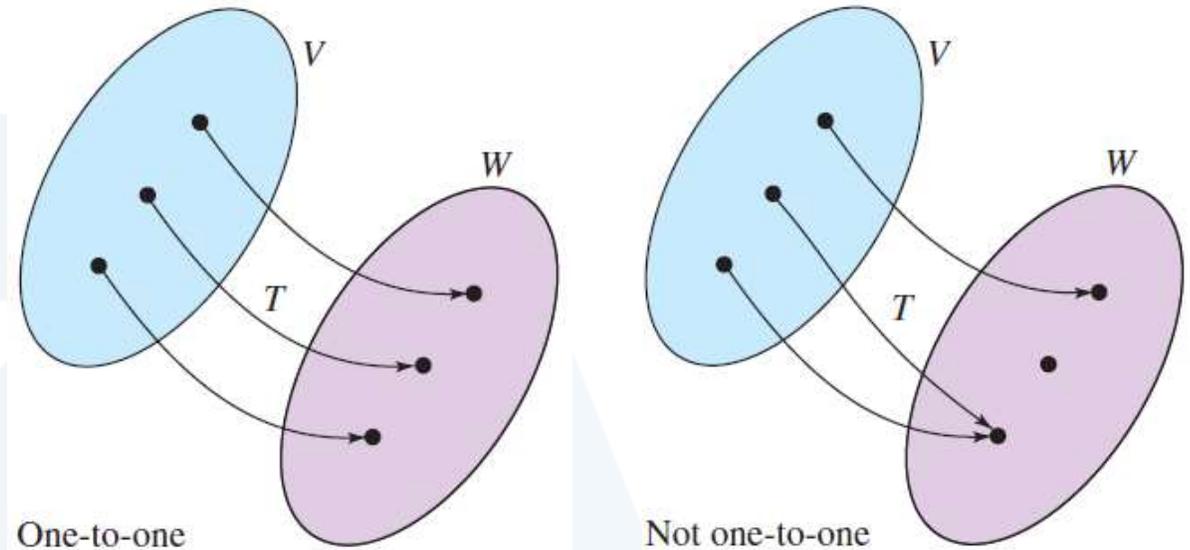
- **Definition:** A function $T: V \rightarrow W$ is **onto** when every element in W has a preimage in V . (T is onto W when W is equal to the range of T).

- **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)**

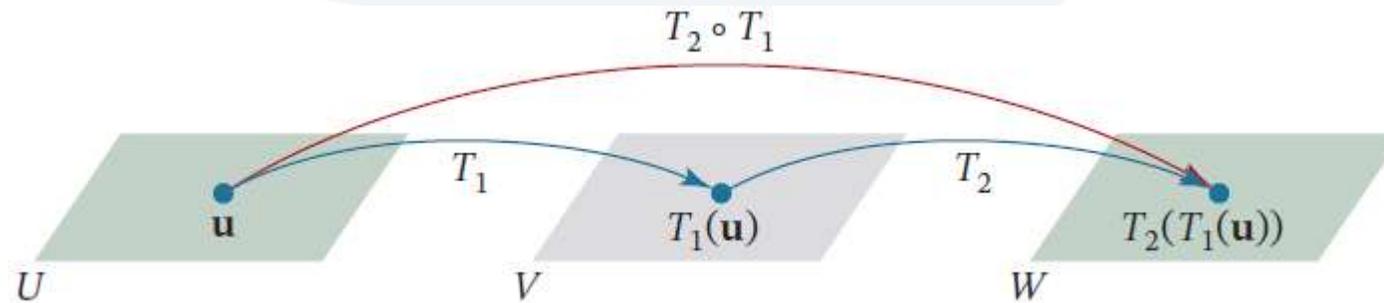
Let $T: R^n \rightarrow R^m$ be a LT given by $T(\mathbf{x}) = A\mathbf{x}$. Find the nullity and rank of T to determine whether T is one-to-one, onto, or neither.

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

$T: R^n \rightarrow R^m$	$\dim(\text{domain of } T)$	$\text{rank}(T)$	$\text{nullity}(T)$	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

(1) The composition $T: R^n \rightarrow R^p$, defined by $T(\mathbf{v}) = T_2(T_1(\mathbf{v}))$, is a LT.

(2) 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:

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

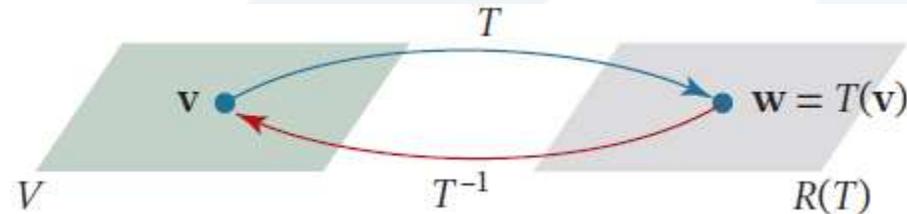
Find the composition $T = T_2 \circ T_1$

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



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

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)

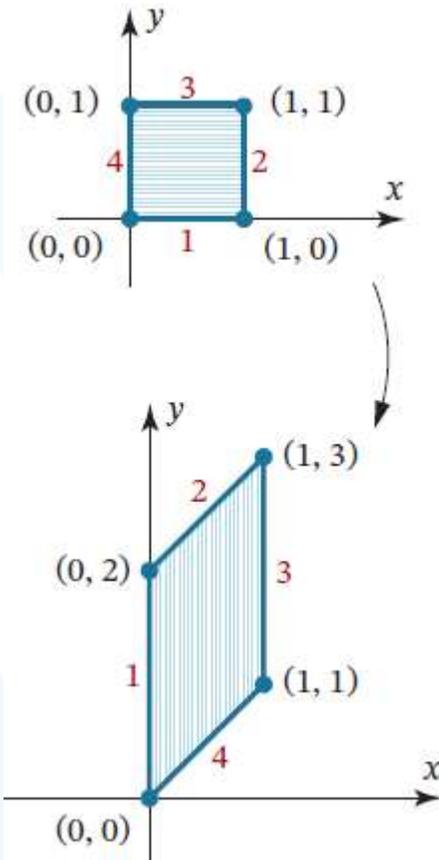
Sketch the image of the unit square under multiplication by the invertible matrix:

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

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

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

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

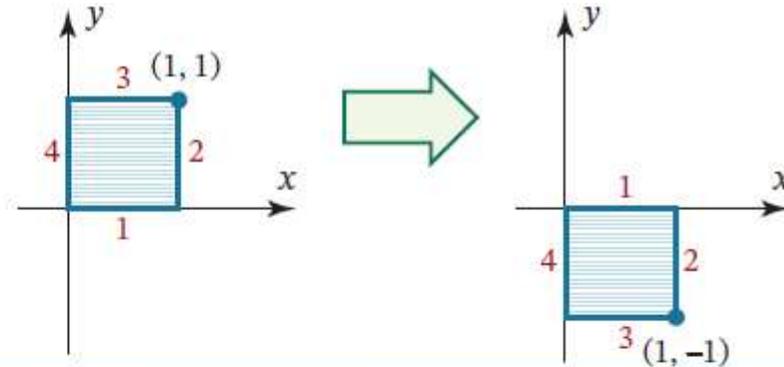




Reflections, Rotations, and Projections

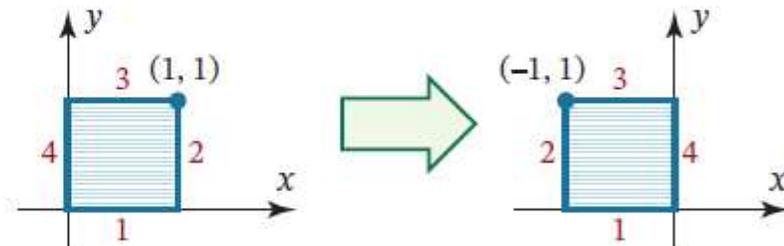
Reflection about the x-axis

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



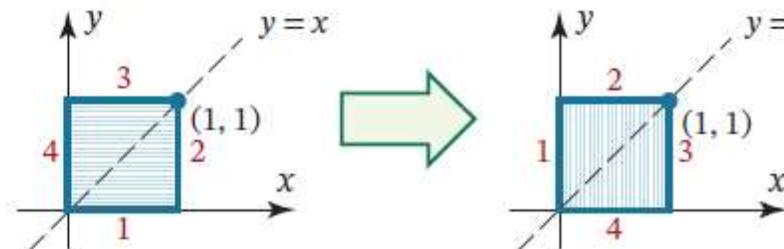
Reflection about the y-axis

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



Reflection about the line $y = x$

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

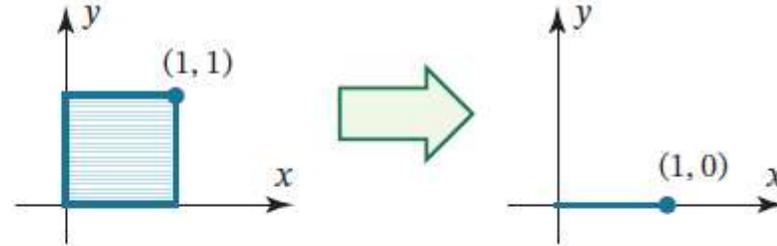




Orthogonal projection onto the x -axis

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

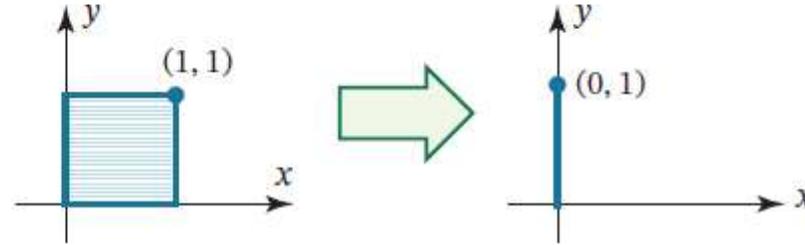
$$T(x, y) = (x, 0)$$



Orthogonal projection onto the y -axis

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

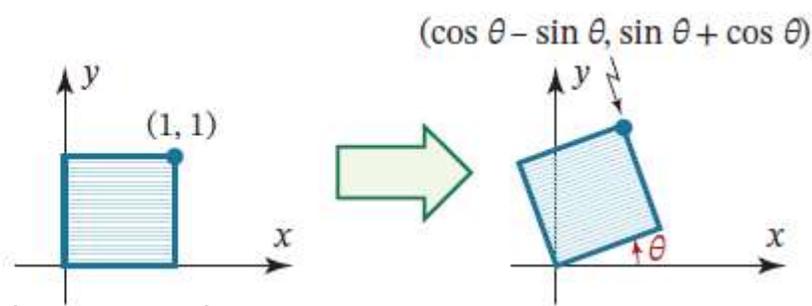
$$T(x, y) = (0, y)$$



Rotation about the origin through a positive angle θ

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

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



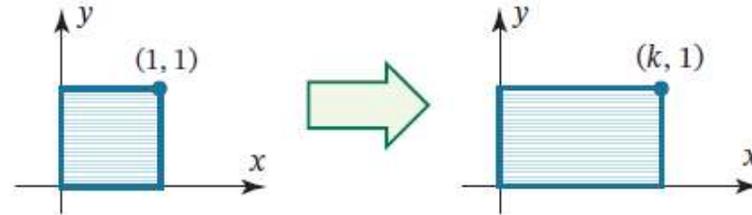


Expansions and Compressions

Expansion in the
x-direction with
factor k

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

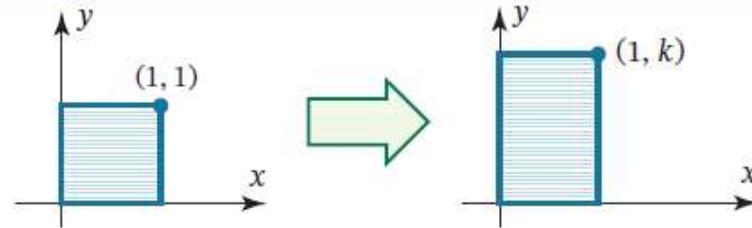
$$(k > 1) \quad T(x, y) = (kx, y)$$



Expansion in the
y-direction with
factor k

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

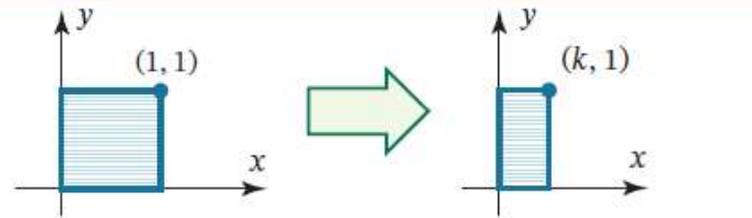
$$(k > 1) \quad T(x, y) = (x, ky)$$



Compression in the
x-direction with
factor k

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

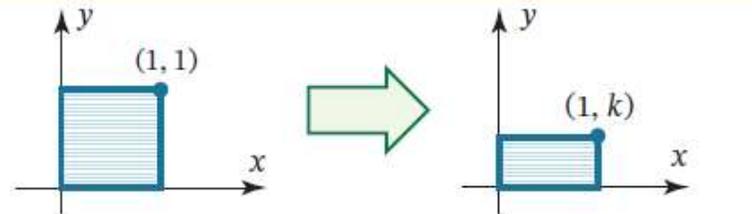
$$(0 < k < 1)$$



Compression in the
y-direction with
factor k

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

$$(0 < k < 1)$$



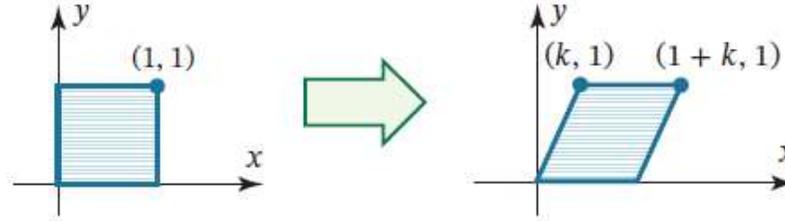


Shears

Shear in the positive x -direction by a factor k

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

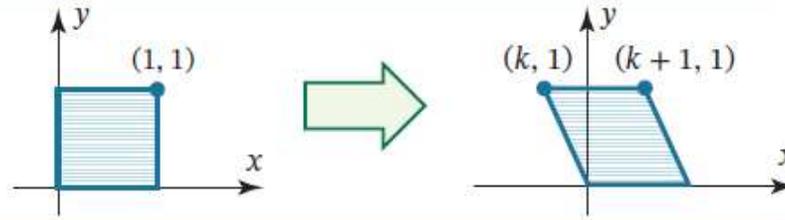
$(k > 0)$ $T(x, y) = (x + ky, y)$



Shear in the negative x -direction by a factor k

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

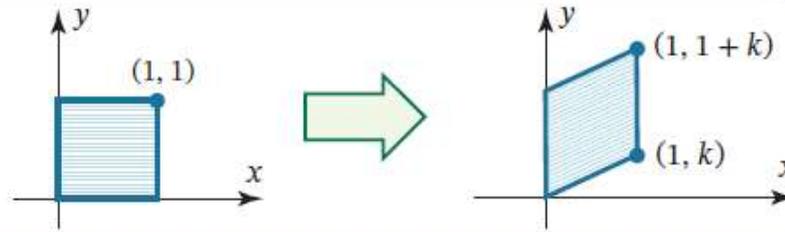
$(k < 0)$



Shear in the positive y -direction by a factor k

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

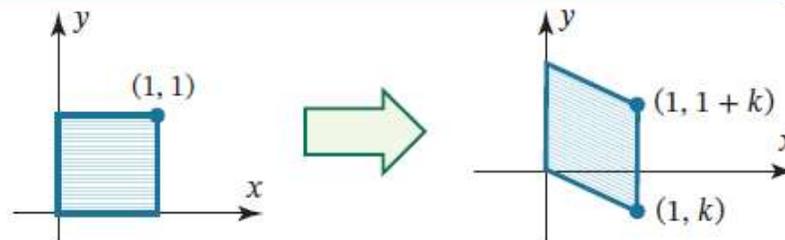
$(k > 0)$ $T(x, y) = (x, kx + y)$



Shear in the negative y -direction by a factor k

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

$(k < 0)$



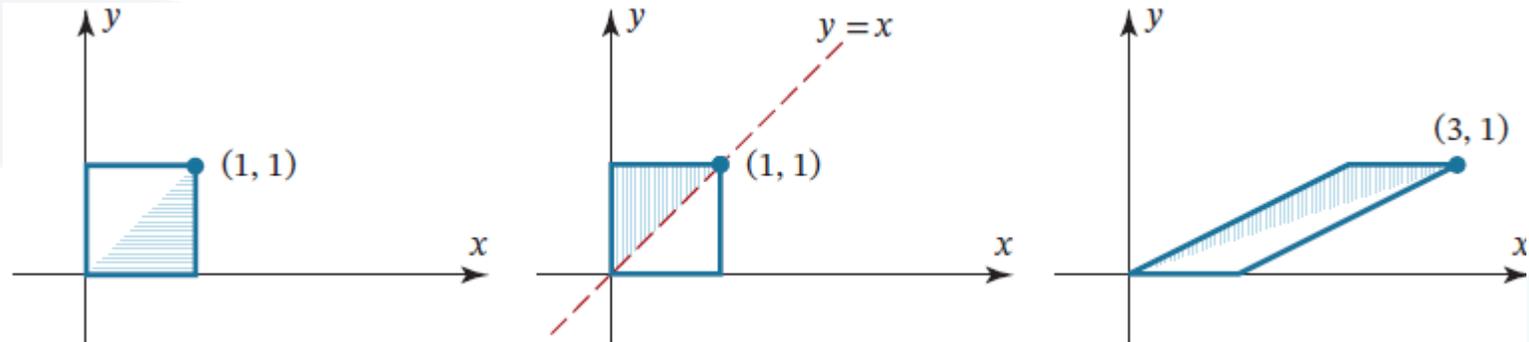


■ **Example 24: (Transformation of the Unit Square)**

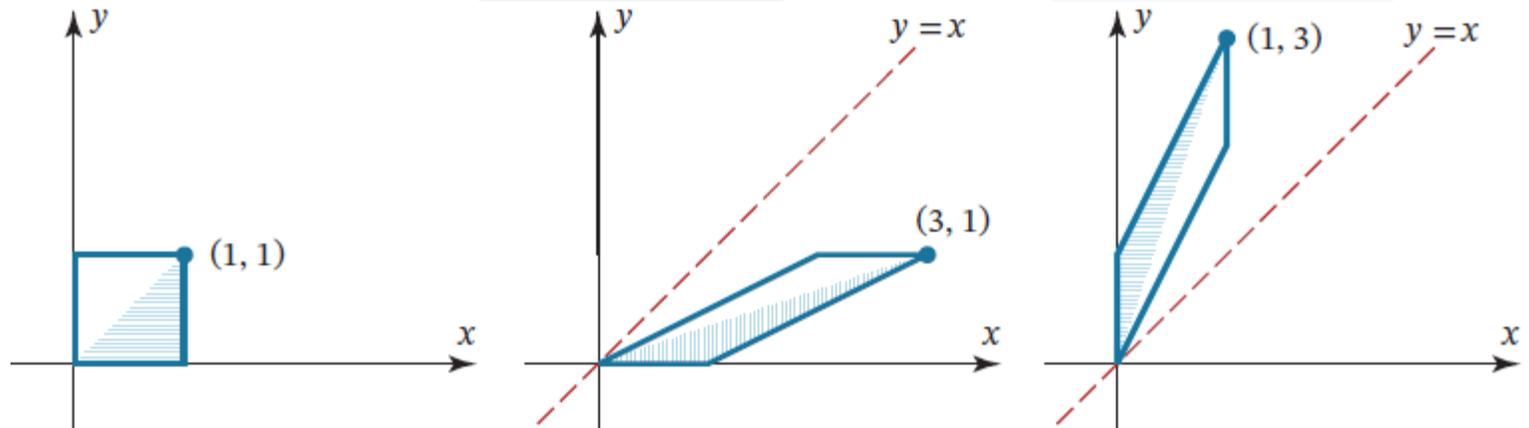
- (a) Find the standard matrix for the operator on R^2 that first shears by a factor of 2 in the x -direction and then reflects the result about the line $y = x$. Sketch the image of the unit square under this operator.
- (b) Find the standard matrix for the operator on R^2 that first reflects about $y = x$ 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 $A_2 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

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



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

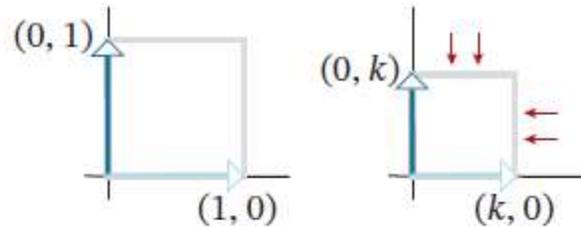


$$A_1 A_2 \neq A_2 A_1$$

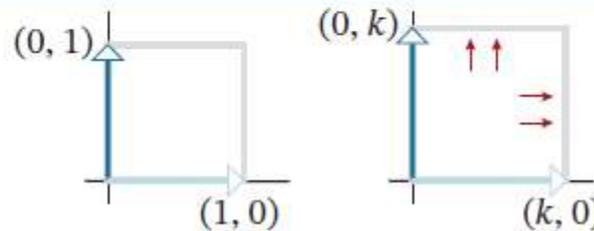


Dilations and Contractions

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:**

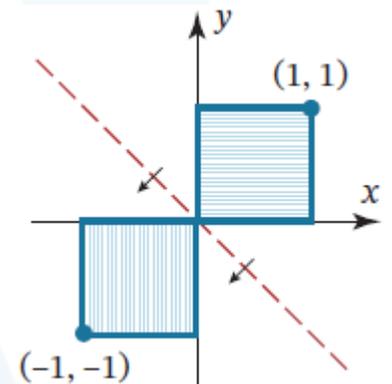
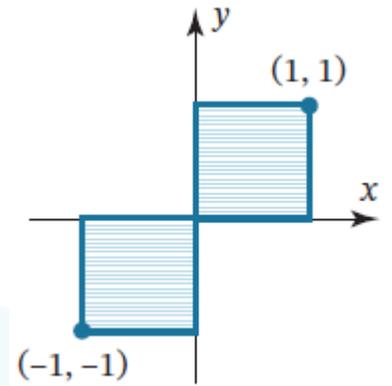
Multiplication by the matrix $A = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}$ has the geometric effect of **reflecting** the unit square about the **origin**.

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

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.

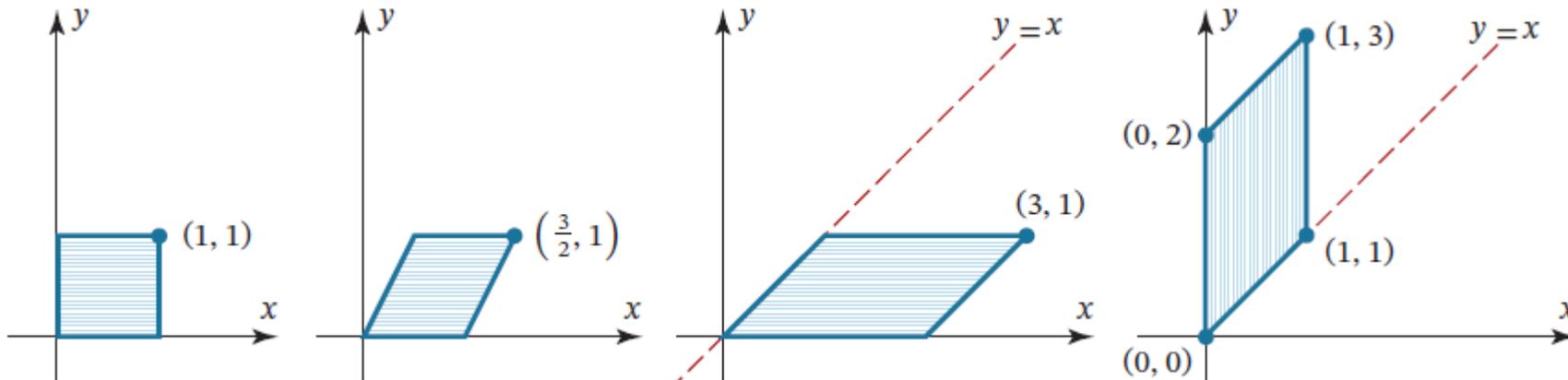
$$A = \begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix} \xrightarrow{r_{12}} \begin{bmatrix} 2 & 1 \\ 0 & 1 \end{bmatrix} \xrightarrow{r_1^{(1/2)}} \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & 1 \end{bmatrix} \xrightarrow{r_{21}^{(-1/2)}} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$$

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

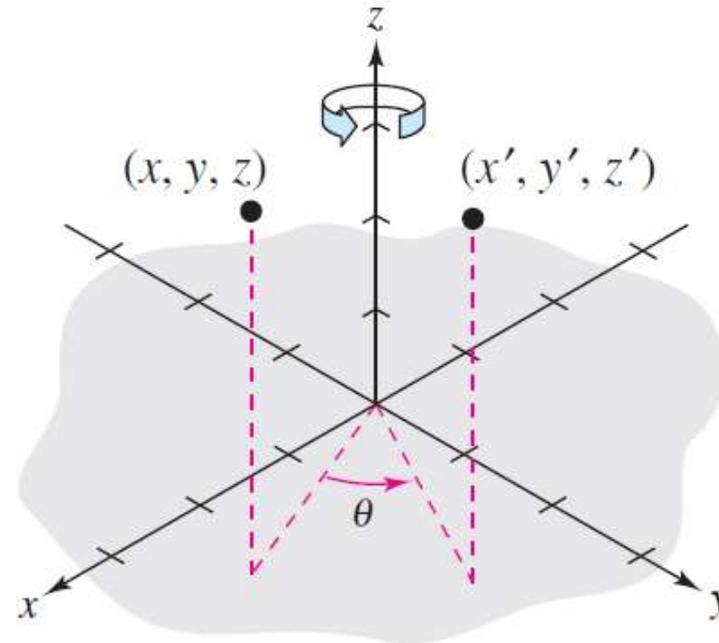


Reading from right to left we can now see that the geometric effect of multiplying by A is equivalent to successively:

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 $y = x$.



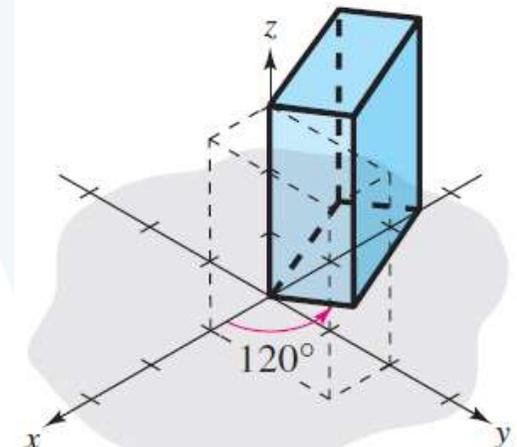
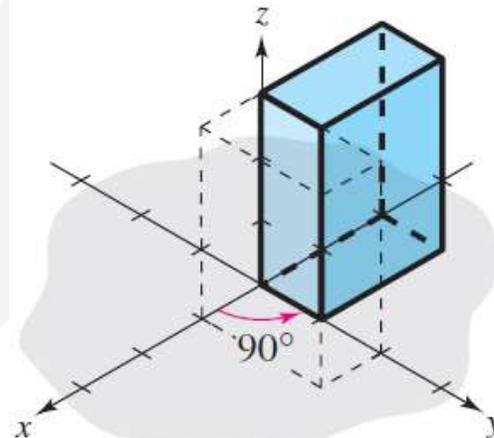
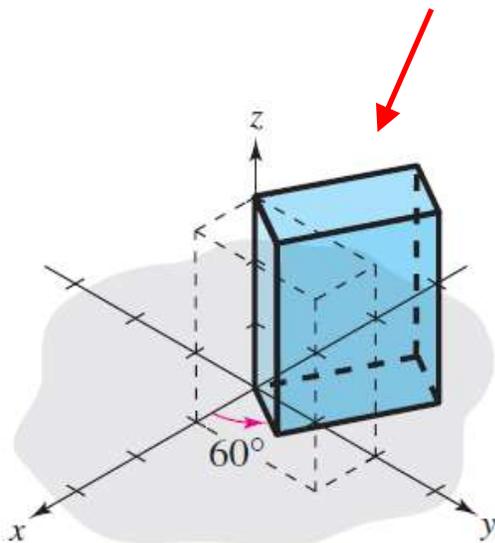
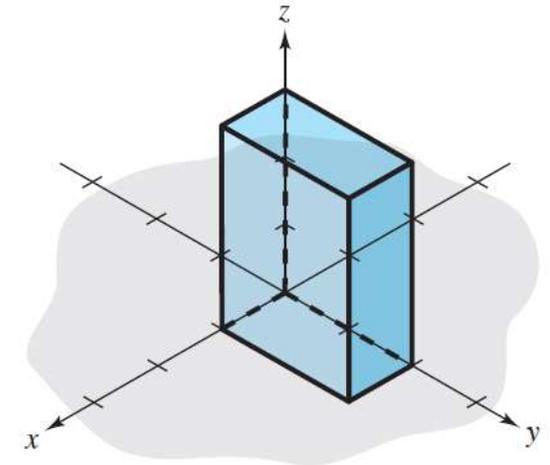
- Rotation In R^3



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

Rotation about the z -axis

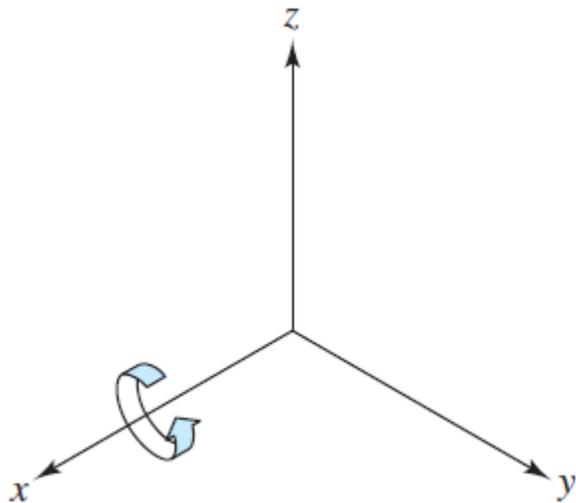
$$A = \begin{bmatrix} \cos 60^\circ & -\sin 60^\circ & 0 \\ \sin 60^\circ & \cos 60^\circ & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1/2 & -\sqrt{3}/2 & 0 \\ \sqrt{3}/2 & 1/2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$





Rotation about the x -axis

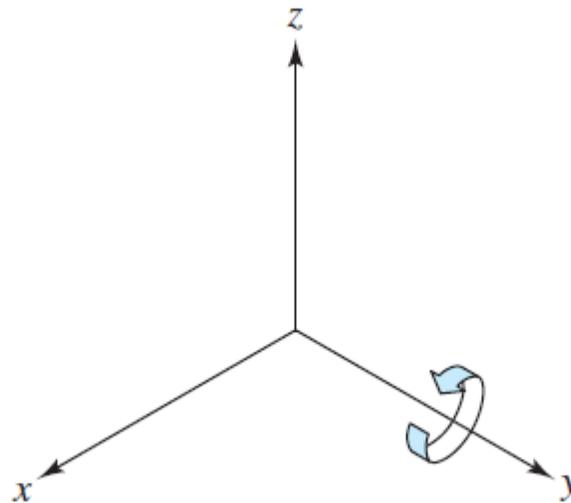
$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \theta & -\sin \theta \\ 0 & \sin \theta & \cos \theta \end{bmatrix}$$



Rotation about x -axis

Rotation about the y -axis

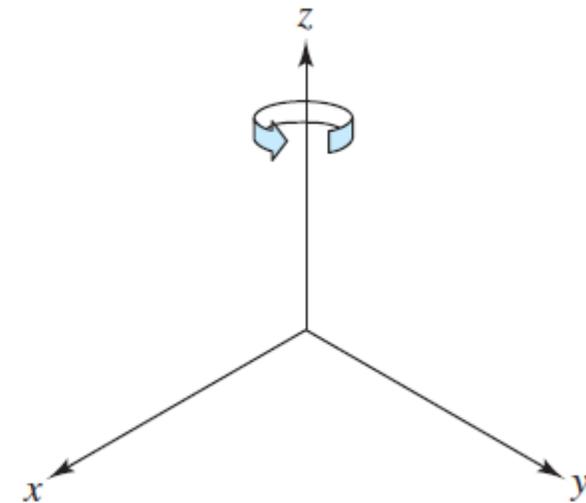
$$\begin{bmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{bmatrix}$$



Rotation about y -axis

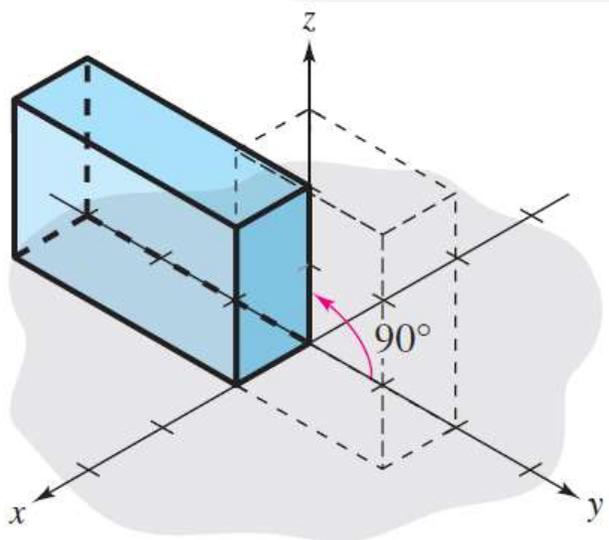
Rotation about the z -axis

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



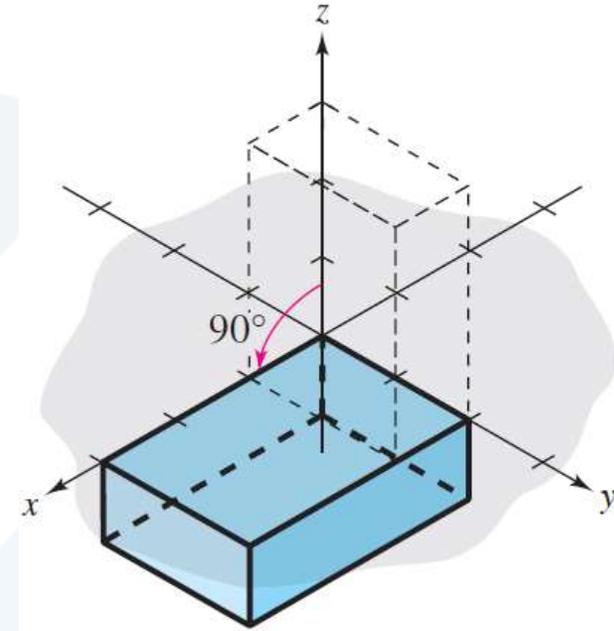
Rotation about z -axis

Rotation of 90° about the x -axis



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

Rotation of 90° about the y -axis



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