

1.9. The Matrix of a Linear Transformation

Theorem. Let $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation. Then there exists a unique matrix A such that

$$T(\bar{x}) = A\bar{x} \quad \text{for all } \bar{x} \text{ in } \mathbb{R}^n.$$

Moreover, if $\bar{e}_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$, $\bar{e}_2 = \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}$, ..., $\bar{e}_n = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$ then

$$A = [T(\bar{e}_1) \ T(\bar{e}_2) \ \dots \ T(\bar{e}_n)].$$

Proof.

$$\text{If } \bar{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} x_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ x_2 \\ \vdots \\ 0 \end{bmatrix} + \dots + \begin{bmatrix} 0 \\ 0 \\ \vdots \\ x_n \end{bmatrix} =$$

$$= x_1 \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix} + \dots + x_n \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} = x_1 \bar{e}_1 + x_2 \bar{e}_2 + \dots + x_n \bar{e}_n.$$

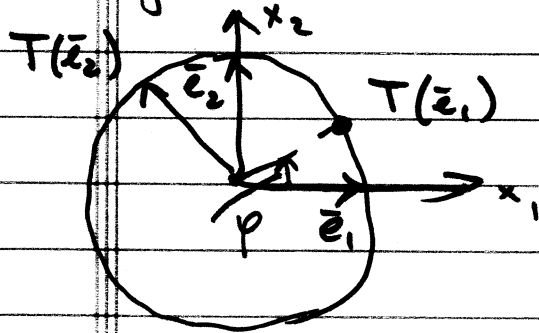
$$\text{Then } T(\bar{x}) = T(x_1 \bar{e}_1 + x_2 \bar{e}_2 + \dots + x_n \bar{e}_n) =$$

$$= x_1 T(\bar{e}_1) + x_2 T(\bar{e}_2) + \dots + x_n T(\bar{e}_n) =$$

$$= [T(\bar{e}_1) \ T(\bar{e}_2) \ \dots \ T(\bar{e}_n)] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = A\bar{x}. //$$

The matrix $A = [T(\bar{e}_1) \ \dots \ T(\bar{e}_n)]$ is called the standard matrix for the linear transformation T .

Ex. Let $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be the transformation that rotates each point in $\mathbb{R}^2$ about the origin through an angle φ , with counterclockwise rotation.



$$T(\vec{e}_1) = (\cos \varphi, \sin \varphi)$$

$$T(\vec{e}_2) = (-\sin \varphi, \cos \varphi)$$

Finally,
$$A = \begin{bmatrix} \cos \varphi & -\sin \varphi \\ \sin \varphi & \cos \varphi \end{bmatrix}.$$

$$T(\vec{x}) = A\vec{x}$$

Definition. A mapping $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is said to be onto $\mathbb{R}^m$ if each $\vec{b}$ in $\mathbb{R}^m$ is the image of at least one $\vec{x}$ in $\mathbb{R}^n$.

A mapping $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is said to be one-to-one if each $\vec{b}$ in $\mathbb{R}^m$ is the image of at most one $\vec{x}$ in $\mathbb{R}^n$.

Ex. Let T be the linear transformation whose standard matrix is

$$A = \begin{bmatrix} 1 & -4 & 8 & 1 \\ 0 & 2 & -1 & 3 \\ 0 & 0 & 0 & 5 \end{bmatrix}.$$

(clear, $\forall \vec{b} \in \mathbb{R}^3$, the equation $A\vec{x} = \vec{b}$ is consistent.

Hence T is onto $\mathbb{R}^3$.

Since the equation $A\vec{x} = \vec{b}$ has a free variable, it is not one-to-one.

Theorem. Let $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation. Then T is one-to-one if and only if the equation $T(\vec{x}) = \vec{0}$ has only the trivial solution.

Theorem. Let $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a linear transformation and let A be the standard matrix for T . Then:

- (a) T maps $\mathbb{R}^n$ onto $\mathbb{R}^m$ if and only if the columns of A span $\mathbb{R}^m$.
- (b) T is one-to-one if and only if the columns of A are linearly independent.

Ex. Let $T(x_1, x_2) = (3x_1 + x_2, 5x_1 + 7x_2, x_1 + 3x_2)$. Show that T is a one-to-one linear transformation.

Sol.
$$T(\vec{x}) = \begin{bmatrix} 3x_1 + x_2 \\ 5x_1 + 7x_2 \\ x_1 + 3x_2 \end{bmatrix} = \begin{bmatrix} 3 & 1 \\ 5 & 7 \\ 1 & 3 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}.$$

Since the matrix columns are linearly independent, T is one-to-one.

TABLE 1 Reflections

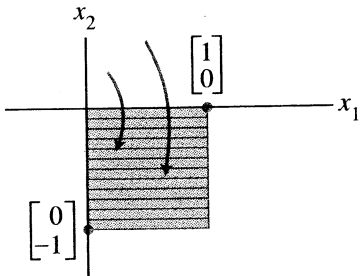
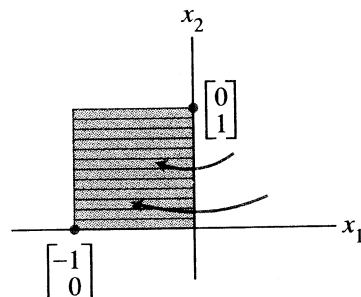
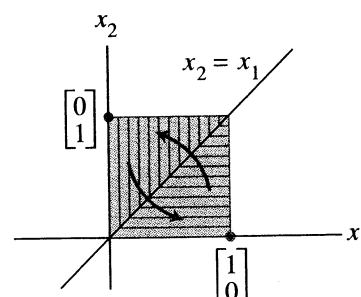
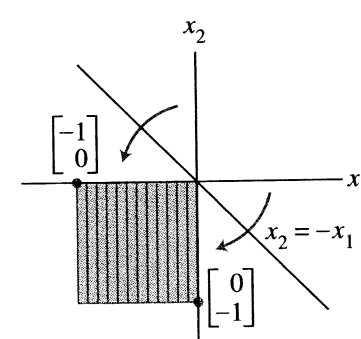
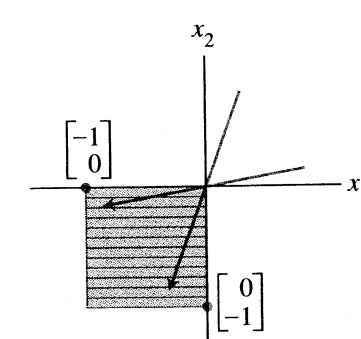
Transformation	Image of the Unit Square	Standard Matrix
Reflection through the x_1 -axis		$\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$
Reflection through the x_2 -axis		$\begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$
Reflection through the line $x_2 = x_1$		$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$
Reflection through the line $x_2 = -x_1$		$\begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$
Reflection through the origin		$\begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}$

TABLE 2 Contractions and Expansions

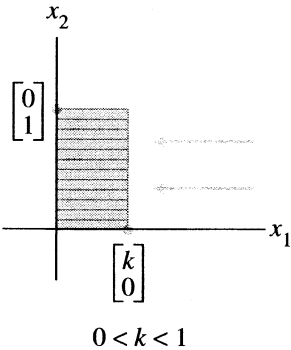
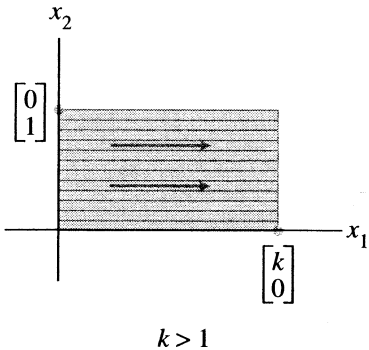
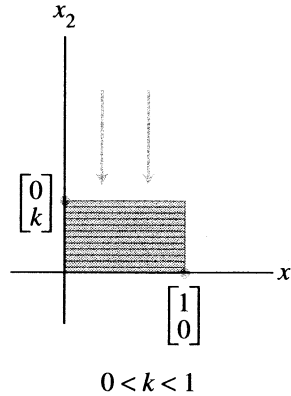
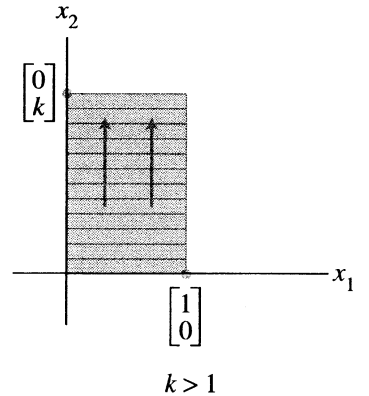
Transformation	Image of the Unit Square		Standard Matrix
Horizontal contraction and expansion			$\begin{bmatrix} k & 0 \\ 0 & 1 \end{bmatrix}$
	$0 < k < 1$	$k > 1$	
Vertical contraction and expansion			$\begin{bmatrix} 1 & 0 \\ 0 & k \end{bmatrix}$
	$0 < k < 1$	$k > 1$	

TABLE 3 Shears

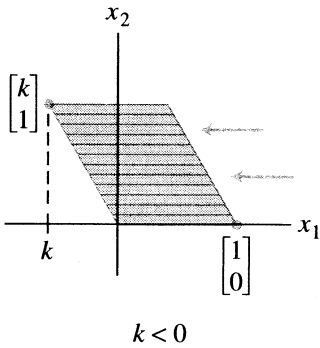
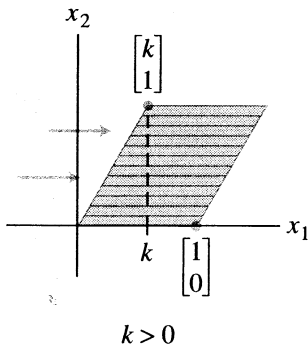
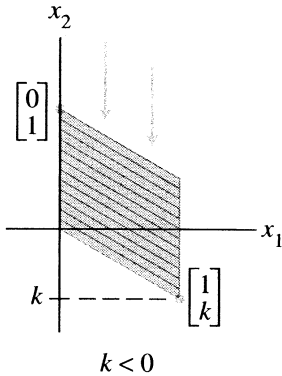
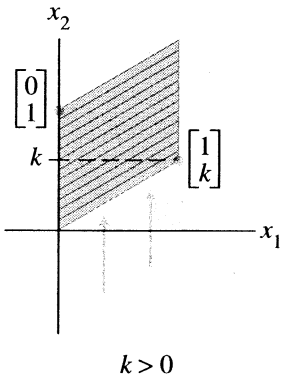
Transformation	Image of the Unit Square		Standard Matrix
Horizontal shear			$\begin{bmatrix} 1 & k \\ 0 & 1 \end{bmatrix}$
	$k < 0$	$k > 0$	
Vertical shear			$\begin{bmatrix} 1 & 0 \\ k & 1 \end{bmatrix}$
	$k < 0$	$k > 0$	

TABLE 4 Projections

Transformation	Image of the Unit Square	Standard Matrix
Projection onto the x_1 -axis	The diagram shows a Cartesian coordinate system with axes x_1 and x_2. The unit square is represented by the vertices $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$, $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$, $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$, and $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$. The projection onto the x_1-axis is the segment from $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$ to $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$. Arrows show the mapping of the top edge to the origin and the right edge to the point $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$.	$\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$
Projection onto the x_2 -axis	The diagram shows a Cartesian coordinate system with axes x_1 and x_2. The unit square is represented by the vertices $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$, $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$, $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$, and $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$. The projection onto the x_2-axis is the segment from $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$ to $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$. Arrows show the mapping of the right edge to the point $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ and the bottom edge to the origin.	$\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$