

## Section 1.9: The Matrix for a Linear Transformation

**Definition:** A transformation  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a **linear transformation** provided for every vector  $\mathbf{u}$  and  $\mathbf{v}$  in  $\mathbb{R}^n$  and every scalar  $c$

$$T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v}), \quad \text{and}$$

$$T(c\mathbf{u}) = cT(\mathbf{u}).$$

**Remark:** We know that a mapping defined by matrix multiplication  $\mathbf{x} \mapsto A\mathbf{x}$  is a linear transformation. In fact, every linear transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  can be realized in terms of matrix multiplication.

## Elementary Vectors

**Elementary Vectors:** We'll use the notation  $\mathbf{e}_i$  to denote the vector in  $\mathbb{R}^n$  having a 1 in the  $i^{\text{th}}$  position and zero everywhere else.

e.g. in  $\mathbb{R}^2$  the elementary vectors are

$$\mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad \text{and} \quad \mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix},$$

in  $\mathbb{R}^3$  they would be

$$\mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \quad \mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad \text{and} \quad \mathbf{e}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

and so forth.

Note that in  $\mathbb{R}^n$ , the elementary vectors are the columns of the identity  $I_n$ .

## Example: Matrix of Linear Transformation

Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^4$  be a linear transformation, and suppose

$$T(\mathbf{e}_1) = \begin{bmatrix} 0 \\ 1 \\ -2 \\ 4 \end{bmatrix}, \quad \text{and} \quad T(\mathbf{e}_2) = \begin{bmatrix} 1 \\ 1 \\ -1 \\ 6 \end{bmatrix}.$$

Use the fact that  $T$  is linear, and the fact that for each  $\mathbf{x}$  in  $\mathbb{R}^2$  we have

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix} = x_1 \mathbf{e}_1 + x_2 \mathbf{e}_2$$

to find a matrix  $A$  such that

$$T(\mathbf{x}) = A\mathbf{x} \quad \text{for every} \quad \mathbf{x} \in \mathbb{R}^2.$$

$$T(\mathbf{e}_1) = \begin{bmatrix} 0 \\ 1 \\ -2 \\ 4 \end{bmatrix}, \quad \text{and} \quad T(\mathbf{e}_2) = \begin{bmatrix} 1 \\ 1 \\ -1 \\ 6 \end{bmatrix}$$

Let  $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  be arbitrary in  $\mathbb{R}^2$

$$\vec{x} = x_1 \vec{e}_1 + x_2 \vec{e}_2. \quad \text{So}$$

$$\begin{aligned} T(\vec{x}) &= T(x_1 \vec{e}_1 + x_2 \vec{e}_2) \\ &= T(x_1 \vec{e}_1) + T(x_2 \vec{e}_2) \\ &= x_1 T(\vec{e}_1) + x_2 T(\vec{e}_2) \end{aligned}$$

*because  
T is linear*

$$= x_1 \begin{bmatrix} 0 \\ 1 \\ -2 \\ 4 \end{bmatrix} + x_2 \begin{bmatrix} 1 \\ -1 \\ -1 \\ 6 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1 \\ 1 & -1 \\ -2 & -1 \\ 4 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

Hence  $T(\vec{x}) = A\vec{x}$  if  $A = \begin{bmatrix} 0 & 1 \\ 1 & -1 \\ -2 & -1 \\ 4 & 6 \end{bmatrix}$ .

Notice that

$$A = [T(\vec{e}_1) \quad T(\vec{e}_2)]$$

# Theorem

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

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

Moreover, the  $j^{\text{th}}$  column of the matrix  $A$  is the vector  $T(\mathbf{e}_j)$ , where  $\mathbf{e}_j$  is the  $j^{\text{th}}$  column of the  $n \times n$  identity matrix  $I_n$ . That is,

$$A = [T(\mathbf{e}_1) \quad T(\mathbf{e}_2) \quad \cdots \quad T(\mathbf{e}_n)].$$

The matrix  $A$  is called the **standard matrix** for the linear transformation  $T$ .

## Example

Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the scaling transformation (contraction or dilation for  $r > 0$ ) defined by

$$T(\mathbf{x}) = r\mathbf{x}, \quad \text{for positive scalar } r.$$

Find the standard matrix for  $T$ .

Calling the matrix  $A$ ,  $A = [T(\vec{e}_1) \ T(\vec{e}_2)]$

$$\vec{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad T(\vec{e}_1) = r\vec{e}_1 = r \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} r \\ 0 \end{bmatrix}$$

$$\vec{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad T(\vec{e}_2) = r\vec{e}_2 = r \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ r \end{bmatrix}$$

Hence  $A = \begin{bmatrix} r & 0 \\ 0 & r \end{bmatrix}$ .

Check:  $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  is any vector in  $\mathbb{R}^2$

$$\text{Then } A\vec{x} = \begin{bmatrix} r & 0 \\ 0 & r \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} rx_1 + 0x_2 \\ 0x_1 + rx_2 \end{bmatrix} = \begin{bmatrix} rx_1 \\ rx_2 \end{bmatrix}$$

$$= r \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = r\vec{x} = T(\vec{x})$$



## Example: Shear Transformation

Find the standard matrix for the linear transformation from  $\mathbb{R}^2 \rightarrow \mathbb{R}^2$  that maps  $\mathbf{e}_2$  to  $\mathbf{e}_2 - \frac{1}{2}\mathbf{e}_1$  and leaves  $\mathbf{e}_1$  unchanged.

Let's call this transformation  $T$ .

$$T(\vec{e}_1) = \vec{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad T(\vec{e}_2) = \vec{e}_2 - \frac{1}{2}\vec{e}_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} - \frac{1}{2} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} -1/2 \\ 1 \end{bmatrix}$$

Calling the matrix  $A$ ,

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

## Example: Shear Transformation

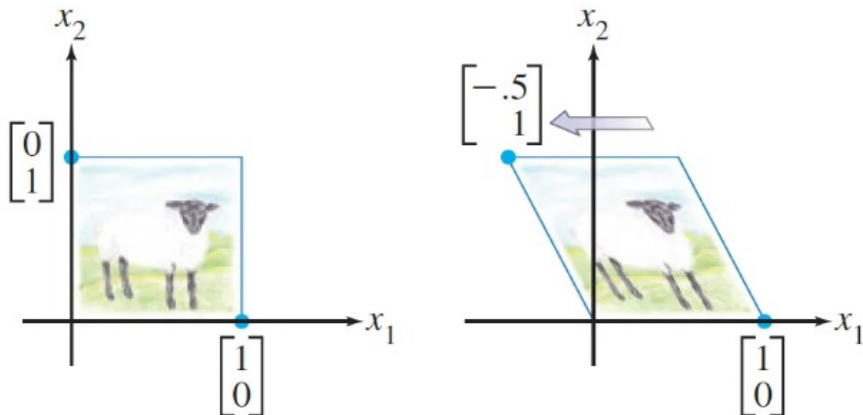
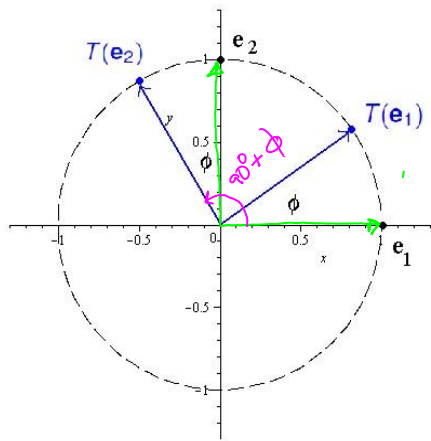


Figure: The unit square under the transformation  $\mathbf{x} \mapsto \begin{bmatrix} 1 & -\frac{1}{2} \\ 0 & 1 \end{bmatrix} \mathbf{x}$ .

## Example: Rotation

Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the rotation transformation that rotates each point in  $\mathbb{R}^2$  counter clockwise about the origin through an angle  $\phi$ . Find the standard matrix for  $T$ .



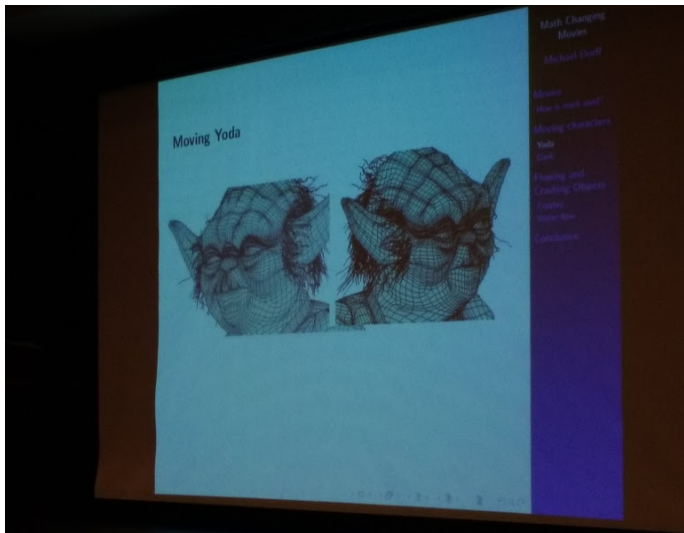
Using some basic trigonometry, the points on the unit circle

$$T(\mathbf{e}_1) = (\cos \phi, \sin \phi)$$

$$\begin{aligned} T(\mathbf{e}_2) &= (\cos(90^\circ + \phi), \sin(90^\circ + \phi)) \\ &= (-\sin \phi, \cos \phi) \end{aligned}$$

$$\text{So } A = \begin{bmatrix} \cos \phi & -\sin \phi \\ \sin \phi & \cos \phi \end{bmatrix}.$$

# Rotation in Animation



# Rotation in Animation

**Moving Yoda**

- ▶ We can move Yoda using matrix multiplication.
- ▶ Store information about the vertices in a  $53756 \times 3$  matrix  $V$ , where row  $i$  of  $V$  contains the  $x$ ,  $y$ , and  $z$  coordinates of the  $i$ th vertex.
- ▶ Yoda can be rotated by  $\theta$  radians about the  $y$ -axis by multiplying  $V$  with  $R$ , where

$$R = \begin{pmatrix} \cos(\theta) & 0 & -\sin(\theta) \\ 0 & 1 & 0 \\ \sin(\theta) & 0 & \cos(\theta) \end{pmatrix}$$

Navigation menu (right side):

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## Example<sup>1</sup>

Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the projection transformation that projects each point onto the  $x_1$  axis

$$T \left( \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \right) = \begin{bmatrix} x_1 \\ 0 \end{bmatrix}.$$

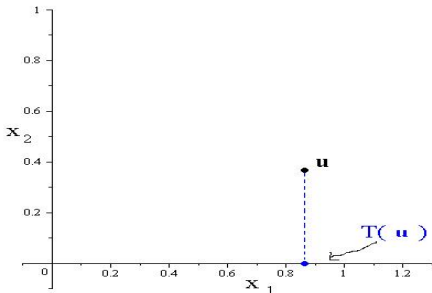
Find the standard matrix for  $T$ .

$$T(\vec{e}_1) = T \left( \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$T(\vec{e}_2) = T \left( \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right) = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Calling the matrix  $A$ ,

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



<sup>1</sup> See pages 77–80 in Lay for matrices associated with other geometric transformation on  $\mathbb{R}^2$

## The Property **Onto**

**Definition:** A mapping  $T : \mathbb{R}^n \longrightarrow \mathbb{R}^m$  is said to be **onto**  $\mathbb{R}^m$  if each  $\mathbf{b}$  in  $\mathbb{R}^m$  is the image of at least one  $\mathbf{x}$  in  $\mathbb{R}^n$ —i.e. if the range of  $T$  is all of the codomain.

If  $T : \mathbb{R}^n \longrightarrow \mathbb{R}^m$  is an **onto** transformation, then the equation

$$T(\mathbf{x}) = \mathbf{b}$$

is always solvable. If  $T$  is a linear transformation with standard matrix  $A$ , then this is equivalent to saying  $A\mathbf{x} = \mathbf{b}$  is always consistent.

Determine if the transformation is onto.

$$T(\mathbf{x}) = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 3 \end{bmatrix} \mathbf{x}.$$

We can state the question as "is the equation  $A\vec{x} = \vec{b}$  always consistent?"

Note that  $A = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 3 \end{bmatrix}$  is  $2 \times 3$ , so

$T: \mathbb{R}^3 \rightarrow \mathbb{R}^2$ . Let  $\vec{b} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$  be any

vector in  $\mathbb{R}^2$ . The augmented matrix



for  $A\vec{x} = \vec{b}$  is  $\begin{bmatrix} 1 & 0 & 2 & b_1 \\ 0 & 1 & 3 & b_2 \end{bmatrix}$ .

This is an rref. The 4th column is not a pivot column for any choice of  $b_1$  and  $b_2$ .

The linear system  $A\vec{x} = \vec{b}$  is always  
consistent. Hence  $T$  is onto.

# The Property **One to One**

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

If  $T : \mathbb{R}^n \longrightarrow \mathbb{R}^m$  is a **one to one** transformation, then the equation

$$T(\mathbf{x}) = T(\mathbf{y}) \quad \text{is only true when} \quad \mathbf{x} = \mathbf{y}.$$