

# February 9 Math 3260 sec. 51 Spring 2024

## Section 1.9: The Matrix for a Linear Transformation

### Recall Linear Transformation

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  $c$

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

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

### Two Remarks

1. Any mapping defined by matrix multiplication,  $\mathbf{x} \mapsto A\mathbf{x}$ , is a linear transformation.
2. Every linear transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  can be realized in terms of matrix multiplication.

# Elementary Vectors

## Definition: 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. The vectors  $\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n$  are called **elementary** vectors.

For example, the elementary vectors in  $\mathbb{R}^2$  are

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

The elementary vectors in  $\mathbb{R}^3$  are

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

# Elementary Vectors

## Remark:

In general, the elementary vectors are the columns of the  $n \times n$  identity matrix.

$$\mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad \mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \quad \dots, \quad \mathbf{e}_n = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

$$I_n = [\mathbf{e}_1 \ \mathbf{e}_2 \ \dots \ \mathbf{e}_n]$$

## Matrix of Linear Transformation: an Example

Suppose  $T : \mathbb{R}^2 \longrightarrow \mathbb{R}^4$  is a linear transformation, and that

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

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

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

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

For  $\vec{x}$  in  $\mathbb{R}^2$ ,  $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_1 \vec{e}_1 + x_2 \vec{e}_2$

$$\begin{aligned} T(\vec{x}) &= T(x_1 \vec{e}_1 + x_2 \vec{e}_2) \\ &= x_1 T(\vec{e}_1) + x_2 T(\vec{e}_2) \\ &= x_1 \begin{bmatrix} 0 \\ 1 \\ -2 \\ 4 \end{bmatrix} + x_2 \begin{bmatrix} 1 \\ 1 \\ -1 \\ 6 \end{bmatrix} \end{aligned}$$

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

call this  
A .

$$\text{Then } T(\vec{x}) = A\vec{x} .$$

# Standard Matrix of a Linear Transformation

## 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  $T$ .

We need  $T(\vec{e}_1)$  and  $T(\vec{e}_2)$  for  $\vec{e}_1, \vec{e}_2$  in  $\mathbb{R}^2$ .

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

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



The standard matrix

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

## A Shear Transformation on $\mathbb{R}^2$

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.

Call the transformation  $S$ .

$$S(\vec{e}_1) = \vec{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\begin{aligned} S(\vec{e}_2) &= \vec{e}_2 - \frac{1}{2}\vec{e}_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} - \frac{1}{2}\begin{bmatrix} 1 \\ 0 \end{bmatrix} \\ &= \begin{bmatrix} -1/2 \\ 1 \end{bmatrix} \end{aligned}$$

The standard matrix

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

## A Shear Transformation on $\mathbb{R}^2$

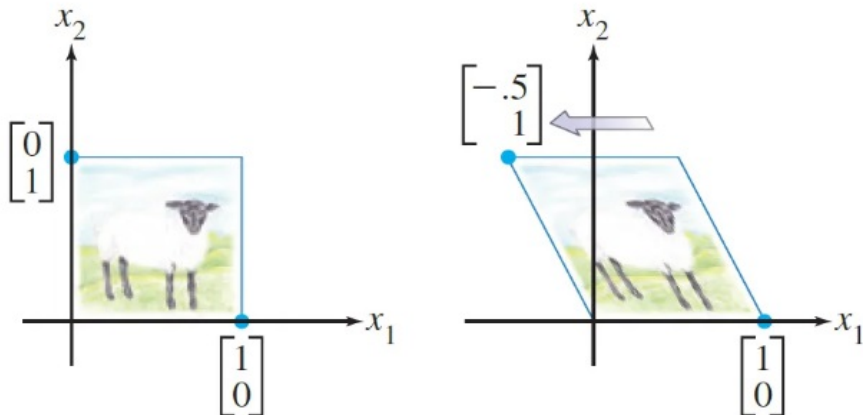
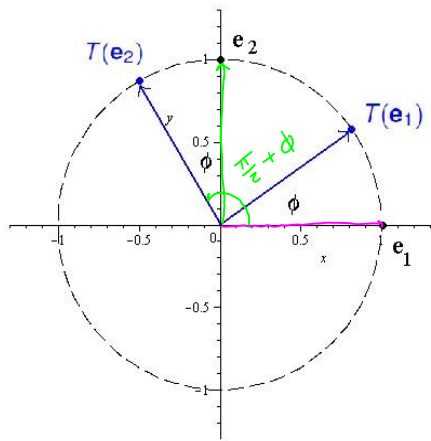


Figure: The unit square under the transformation  $\mathbf{x} \mapsto \begin{bmatrix} 1 & -\frac{1}{2} \\ 0 & 1 \end{bmatrix} \mathbf{x}$ .

## A Rotation on $\mathbb{R}^2$

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  $T$ .



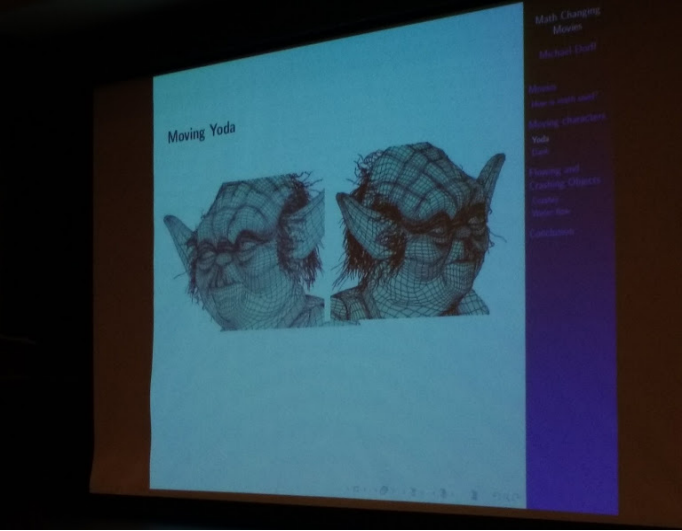
Using some basic trigonometry, the points on the unit circle

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

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

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

# Rotation in Animation



# Rotation in Animation

**Moving Yoda**

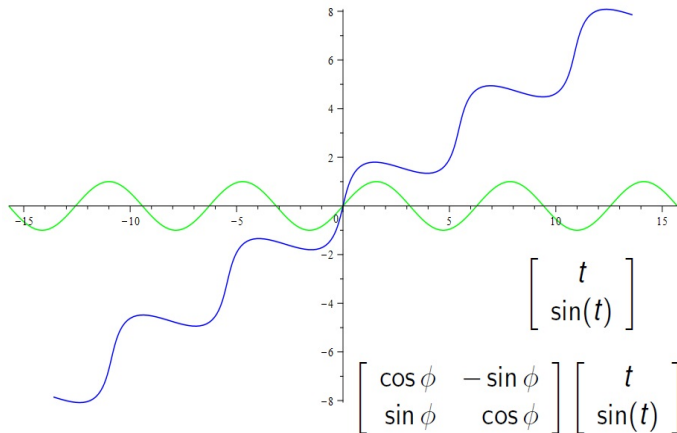
- ▶ We can move Yoda using matrix multiplication.
- ▶ Store information about the vertices in a  $53756 \times 3$  matrix  $V$ , where row  $i$  of  $V$  contains the  $x$ ,  $y$ , and  $z$  coordinates of the  $i$ th vertex.
- ▶ Yoda can be rotated by  $\theta$  radians about the  $y$ -axis by multiplying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

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

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## Rotation in Curve Generation



**Figure:** The curve  $y = \sin(x)$  plotted as a vector valued function along with a version rotated through an angle  $\phi = \frac{\pi}{6}$ .



# Onto and One to One

## Definition

A mapping  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  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.

## Definition

A mapping  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  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  $\mathbb{R}^n$ .

## Some Theorems about *Onto* and *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 homogeneous equation  $T(\mathbf{x}) = \mathbf{0}$  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

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

## Remarks

Suppose  $T : \mathbb{R}^n \longrightarrow \mathbb{R}^m$  is a linear transformation and  $A$  is the standard matrix for  $T$ .

- ▶ If  $T$  is **onto**, then
  - ▶ the range of  $T$  is  $\mathbb{R}^m$ ,
  - ▶ the equation  $T(\mathbf{x}) = \mathbf{b}$  is always solvable,
  - ▶ the system  $A\mathbf{x} = \mathbf{b}$  is always consistent.
  
- ▶ If  $T$  is **one to one**, then
  - ▶  $T(\mathbf{x}) = T(\mathbf{y})$  implies that  $\mathbf{x} = \mathbf{y}$ ,
  - ▶  $A\mathbf{x} = \mathbf{0}$  has no free variables.

## Example

Consider the linear transformation  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$   
 $(x_1, x_2, x_3) \mapsto (x_3, x_1 + x_2)$

Determine the set of all preimages<sup>1</sup> of  $\mathbf{0}$ . State the solution as a span.

$\vec{x}$  is a preimage of  $\vec{0}$  if  $T(\vec{x}) = \vec{0}$

We can use the standard matrix  $A$ .

$$A = [T(\vec{e}_1) \quad T(\vec{e}_2) \quad T(\vec{e}_3)]$$

$$T(\vec{e}_1) = T(1, 0, 0) = (0, 1+0) = (0, 1)$$

$$T(\vec{e}_2) = T(0, 1, 0) = (0, 0+1) = (0, 1)$$

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<sup>1</sup>This actually has a special name. The set of all preimages of the zero vector is called the *kernel* of  $T$ .

$$T(\vec{e}_3) = T(0, 0, 1) = (1, 0+0) = (1, 0)$$

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

$$\text{For } A\vec{x} = \vec{0}, \quad A \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\begin{aligned} x_1 &= -x_2 \\ x_2 & \text{ free} \\ x_3 &= 0 \end{aligned} \quad \vec{x} = \begin{bmatrix} -x_2 \\ x_2 \\ 0 \end{bmatrix} = x_2 \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$$

The set of preimages of  $\vec{0}$  is  
 $\text{Span} \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} \right\}$ .