

## Section 1.9

### The Matrix of a Linear Transformation

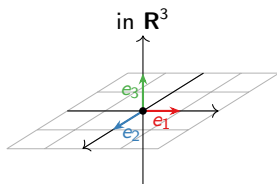
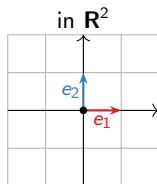
# Unit Coordinate Vectors

## Definition

The **unit coordinate vectors** in  $\mathbf{R}^n$  are

This is what  $e_1, e_2, \dots$  mean,  
for the rest of the class.

$$e_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \\ 0 \end{pmatrix}, \quad e_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \\ 0 \end{pmatrix}, \quad \dots, \quad e_{n-1} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ 0 \end{pmatrix}, \quad e_n = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix}.$$



**Note:** if  $A$  is an  $m \times n$  matrix with columns  $v_1, v_2, \dots, v_n$ , then  $Ae_i = v_i$  for  $i = 1, 2, \dots, n$ : multiplying a matrix by  $e_i$  gives you the  $i$ th column.

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 4 \\ 7 \end{pmatrix} \quad \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 \\ 5 \\ 8 \end{pmatrix} \quad \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 3 \\ 6 \\ 9 \end{pmatrix}$$

# Linear Transformations are Matrix Transformations

**Recall:** A matrix  $A$  defines a linear transformation  $T$  by  $T(x) = Ax$ .

## Theorem

Let  $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$  be a linear transformation. Let

$$A = \begin{pmatrix} \left| \begin{array}{c} T(e_1) \\ \vdots \end{array} \right| & \left| \begin{array}{c} T(e_2) \\ \vdots \end{array} \right| & \cdots & \left| \begin{array}{c} T(e_n) \\ \vdots \end{array} \right| \end{pmatrix}.$$

This is an  $m \times n$  matrix, and  $T$  is the matrix transformation for  $A$ :  $T(x) = Ax$ .

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

### Take-Away

Linear transformations are the same as matrix transformations.

## Dictionary

Linear transformation  $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$   $\rightsquigarrow$   $m \times n$  matrix  $A = \begin{pmatrix} \left| \begin{array}{c} T(e_1) \\ \vdots \end{array} \right| & \left| \begin{array}{c} T(e_2) \\ \vdots \end{array} \right| & \cdots & \left| \begin{array}{c} T(e_n) \\ \vdots \end{array} \right| \end{pmatrix}$

$T(x) = Ax$   $\longleftarrow$   $m \times n$  matrix  $A$

$T: \mathbf{R}^n \rightarrow \mathbf{R}^m$

# Linear Transformations are Matrix Transformations

Continued

Why is a linear transformation a matrix transformation?

Suppose for simplicity that  $T: \mathbf{R}^3 \rightarrow \mathbf{R}^2$ .

$$\begin{aligned} T \begin{pmatrix} x \\ y \\ z \end{pmatrix} &= T \left( x \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + y \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + z \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \right) \\ &= T(x\mathbf{e}_1 + y\mathbf{e}_2 + z\mathbf{e}_3) \\ &= xT(\mathbf{e}_1) + yT(\mathbf{e}_2) + zT(\mathbf{e}_3) \\ &= \begin{pmatrix} | & | & | \\ T(\mathbf{e}_1) & T(\mathbf{e}_2) & T(\mathbf{e}_3) \\ | & | & | \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} \\ &= A \begin{pmatrix} x \\ y \\ z \end{pmatrix}. \end{aligned}$$

# Linear Transformations are Matrix Transformations

## Example

Before, we defined a **dilation** transformation  $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$  by  $T(x) = 1.5x$ .  
What is its standard matrix?

$$\left. \begin{aligned} T(e_1) &= 1.5e_1 = \begin{pmatrix} 1.5 \\ 0 \end{pmatrix} \\ T(e_2) &= 1.5e_2 = \begin{pmatrix} 0 \\ 1.5 \end{pmatrix} \end{aligned} \right\} \Rightarrow A = \begin{pmatrix} 1.5 & 0 \\ 0 & 1.5 \end{pmatrix}.$$

Check:

$$\begin{pmatrix} 1.5 & 0 \\ 0 & 1.5 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 1.5x \\ 1.5y \end{pmatrix} = 1.5 \begin{pmatrix} x \\ y \end{pmatrix} = T \begin{pmatrix} x \\ y \end{pmatrix}.$$

# Linear Transformations are Matrix Transformations

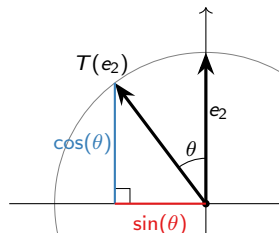
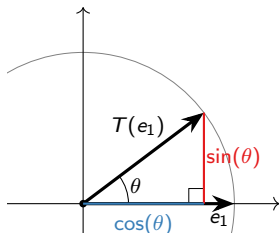
## Example

### Question

What is the matrix for the linear transformation  $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$  defined by

$$T(x) = x \text{ rotated counterclockwise by an angle } \theta?$$

(Check linearity...)



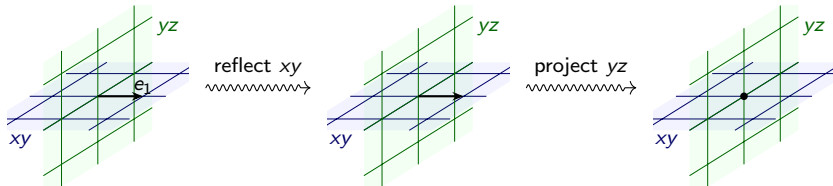
$$\left. \begin{aligned} T(e_1) &= \begin{pmatrix} \cos(\theta) \\ \sin(\theta) \end{pmatrix} \\ T(e_2) &= \begin{pmatrix} -\sin(\theta) \\ \cos(\theta) \end{pmatrix} \end{aligned} \right\} \Rightarrow A = \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix} \quad \left( \begin{array}{l} \theta = 90^\circ \Rightarrow \\ A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \\ \text{(from before)} \end{array} \right)$$

# Linear Transformations are Matrix Transformations

## Example

### Question

What is the matrix for the linear transformation  $T: \mathbf{R}^3 \rightarrow \mathbf{R}^3$  that reflects through the  $xy$ -plane and then projects onto the  $yz$ -plane?



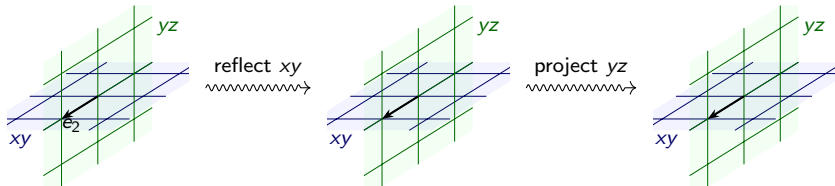
$$T(e_1) = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}.$$

# Linear Transformations are Matrix Transformations

Example, continued

## Question

What is the matrix for the linear transformation  $T: \mathbf{R}^3 \rightarrow \mathbf{R}^3$  that reflects through the  $xy$ -plane and then projects onto the  $yz$ -plane?



$$T(e_2) = e_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}.$$

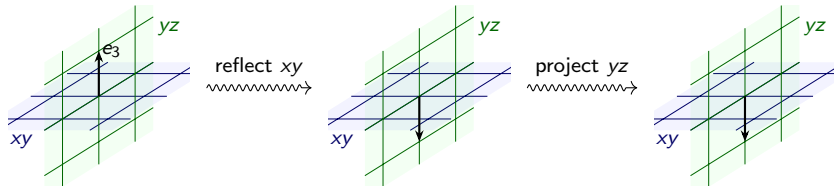


# Linear Transformations are Matrix Transformations

Example, continued

## Question

What is the matrix for the linear transformation  $T: \mathbf{R}^3 \rightarrow \mathbf{R}^3$  that reflects through the  $xy$ -plane and then projects onto the  $yz$ -plane?



$$T(e_3) = \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}.$$

# Linear Transformations are Matrix Transformations

Example, continued

## Question

What is the matrix for the linear transformation  $T: \mathbf{R}^3 \rightarrow \mathbf{R}^3$  that reflects through the  $xy$ -plane and then projects onto the  $yz$ -plane?

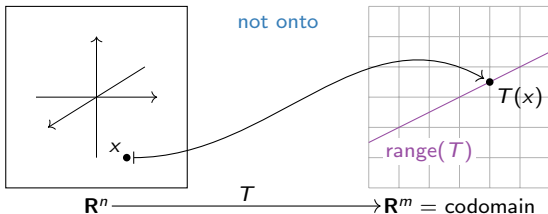
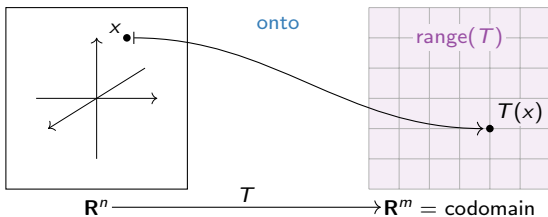
$$\left. \begin{aligned} T(e_1) &= \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \\ T(e_2) &= \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \\ T(e_3) &= \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix} \end{aligned} \right\} \implies A = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix}.$$

There is a long list of geometric transformations of  $\mathbf{R}^2$  in §1.9 of Lay. (Reflections over the diagonal, contractions and expansions along different axes, shears, projections, ...) Please look them over.

# Onto Transformations

## Definition

A transformation  $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$  is **onto** (or **surjective**) if the range of  $T$  is equal to  $\mathbf{R}^m$  (its codomain). In other words, each  $b$  in  $\mathbf{R}^m$  is the image of *at least one*  $x$  in  $\mathbf{R}^n$ : every possible output has an input. Note that *not* onto means there is some  $b$  in  $\mathbf{R}^m$  which is not the image of any  $x$  in  $\mathbf{R}^n$ .



# Characterization of Onto Transformations

## Theorem

Let  $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$  be a linear transformation with matrix  $A$ . Then the following are equivalent:

- ▶  $T$  is onto
- ▶  $T(x) = b$  has a solution for every  $b$  in  $\mathbf{R}^m$
- ▶  $Ax = b$  is consistent for every  $b$  in  $\mathbf{R}^m$
- ▶ The columns of  $A$  span  $\mathbf{R}^m$
- ▶  $A$  has a pivot in every row

## Question

If  $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$  is onto, what can we say about the relative sizes of  $n$  and  $m$ ?

**Answer:**  $T$  corresponds to an  $m \times n$  matrix  $A$ . In order for  $A$  to have a pivot in every row, it must have *at least as many* columns as rows:  $m \leq n$ .

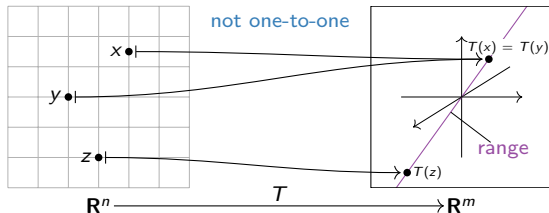
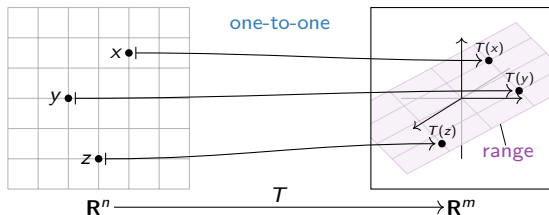
$$\begin{pmatrix} \color{red}{1} & 0 & \star & 0 & \star \\ 0 & \color{red}{1} & \star & 0 & \star \\ 0 & 0 & 0 & \color{red}{1} & \star \end{pmatrix}$$

For instance,  $\mathbf{R}^2$  is “too small” to map *onto*  $\mathbf{R}^3$ .

# One-to-one Transformations

## Definition

A transformation  $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$  is **one-to-one** (or **into**, or **injective**) if different vectors in  $\mathbf{R}^n$  map to different vectors in  $\mathbf{R}^m$ . In other words, each  $b$  in  $\mathbf{R}^m$  is the image of *at most one*  $x$  in  $\mathbf{R}^n$ : different inputs have different outputs. Note that *not* one-to-one means different vectors in  $\mathbf{R}^n$  have the same image.



# Characterization of One-to-One Transformations

## Theorem

Let  $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$  be a linear transformation with matrix  $A$ . Then the following are equivalent:

- ▶  $T$  is one-to-one
- ▶  $T(x) = b$  has one or zero solutions for every  $b$  in  $\mathbf{R}^m$
- ▶  $Ax = b$  has a unique solution or is inconsistent for every  $b$  in  $\mathbf{R}^m$
- ▶  $Ax = 0$  has a unique solution
- ▶ The columns of  $A$  are linearly independent
- ▶  $A$  has a pivot in every column.

## Question

If  $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$  is one-to-one, what can we say about the relative sizes of  $n$  and  $m$ ?

**Answer:**  $T$  corresponds to an  $m \times n$  matrix  $A$ . In order for  $A$  to have a pivot in every column, it must have *at least as many rows as columns*:  $n \leq m$ .

$$\begin{pmatrix} \color{red}{1} & 0 & 0 \\ 0 & \color{red}{1} & 0 \\ 0 & 0 & \color{red}{1} \\ 0 & 0 & 0 \end{pmatrix}$$

For instance,  $\mathbf{R}^3$  is “too big” to map *into*  $\mathbf{R}^2$ .