

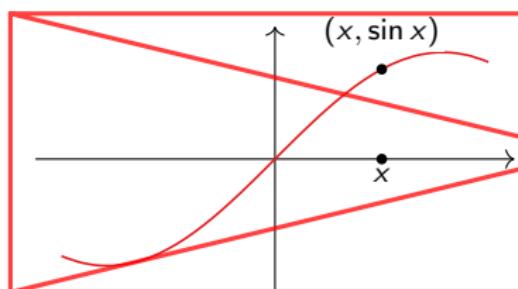
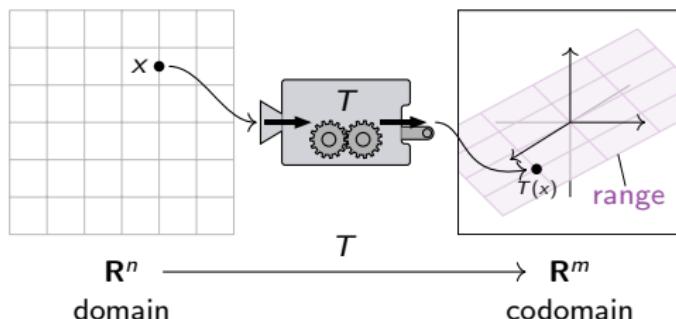
- ▶ Please complete the mid-semester CIOS survey this week.
- ▶ The first midterm will take place during recitation **a week from Friday**, September 30. It covers Chapter 1, sections 1–5 and 7–9.
- ▶ Homeworks 1.5, 1.7, 1.8 are due Friday.
  - ▶ There are three this week so that there can be two next week, the week of the midterm.
- ▶ Quiz on Friday: sections 1.5 and 1.7.
- ▶ My office hours are Wednesday, 1–2pm and Thursday, 3:30–4:30pm, in Skiles 221.
  - ▶ I'll have extra office hours next week.
  - ▶ As always, TAs' office hours are posted on the website.
  - ▶ Also there are links to other resources like Math Lab.

# Transformations

Review from last time

## Definition

A **transformation** (or **function** or **map**) from  $\mathbf{R}^n$  to  $\mathbf{R}^m$  is a rule  $T$  that assigns to each vector  $x$  in  $\mathbf{R}^n$  a vector  $T(x)$  in  $\mathbf{R}^m$ .



You may be used to thinking of a function in terms of its graph. But for a function from  $\mathbf{R}^m$  to  $\mathbf{R}^n$ , the graph needs  $m + n$  dimensions, so it's hard to visualize and draw. We won't be graphing most of the transformations in this class.

## Matrix Transformations

Most of the transformations we encounter in this class will come from a matrix.

### Definition

Let  $A$  be an  $m \times n$  matrix. The **matrix transformation** associated to  $A$  is the transformation

$$T: \mathbf{R}^n \longrightarrow \mathbf{R}^m \quad \text{defined by} \quad T(x) = Ax.$$

In other words,  $T$  takes the vector  $x$  in  $\mathbf{R}^n$  to the vector  $Ax$  in  $\mathbf{R}^m$ .

- ▶ The *domain* of  $T$  is  $\mathbf{R}^n$ , which is the number of columns of  $A$ .
- ▶ The *codomain* of  $T$  is  $\mathbf{R}^m$ , which is the number of rows of  $A$ .
- ▶ The *range* of  $T$  is the set of all images of  $T$ :

$$T(x) = Ax = \begin{pmatrix} | & | & & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & & | \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = x_1 v_1 + x_2 v_2 + \cdots + x_n v_n.$$

This is the column span of  $A$ .

## Matrix Transformations

### Example

Let  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 1 \end{pmatrix}$  and let  $T(x) = Ax$ , so  $T: \mathbf{R}^2 \rightarrow \mathbf{R}^3$ .

- If  $u = \begin{pmatrix} 3 \\ 4 \end{pmatrix}$  then  $T(u) = \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 7 \\ 4 \\ 7 \end{pmatrix}$ .
- Let  $b = \begin{pmatrix} 7 \\ 5 \\ 7 \end{pmatrix}$ . Find  $v$  in  $\mathbf{R}^2$  such that  $T(v) = b$ . Is there more than one?

We want to find  $v$  such that  $Av = b$ . We know how to do that:

$$\begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 1 \end{pmatrix} v = \begin{pmatrix} 7 \\ 5 \\ 7 \end{pmatrix} \xrightarrow{\text{augmented matrix}} \left( \begin{array}{cc|c} 1 & 1 & 7 \\ 0 & 1 & 5 \\ 1 & 1 & 7 \end{array} \right) \xrightarrow{\text{row reduce}} \left( \begin{array}{cc|c} 1 & 0 & 2 \\ 0 & 1 & 5 \\ 0 & 0 & 0 \end{array} \right).$$

This gives  $x = 2$  and  $y = 5$ , or  $v = \begin{pmatrix} 2 \\ 5 \end{pmatrix}$  (unique). In other words,

$$T(v) = \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 2 \\ 5 \end{pmatrix} = \begin{pmatrix} 7 \\ 5 \\ 7 \end{pmatrix}.$$

## Matrix Transformations

Example, continued

Let  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 1 \end{pmatrix}$  and let  $T(x) = Ax$ , so  $T: \mathbf{R}^2 \rightarrow \mathbf{R}^3$ .

- ▶ Is there any  $c$  in  $\mathbf{R}^3$  such that there is more than one  $w \in \mathbf{R}^2$  with  $T(w) = c$ ?

**Translation:** is there any  $c$  in  $\mathbf{R}^3$  such that the solution set for  $Ax = c$  has more than one vector  $w$  in it?

The solution set to  $Ax = b$  has only one vector  $v$ . This is a translate of the solution set to  $Ax = 0$ . So is the solution set to  $Ax = c$ . So no!

- ▶ Find  $c$  such that there is *no*  $v$  with  $T(v) = c$ .

**Translation:** Find  $c$  such that  $Ax = c$  is inconsistent.

**Translation:** Find  $c$  not in the column span of  $A$  (i.e., the range of  $T$ ).

We could draw a picture, or notice:  $a \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} + b \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} a+b \\ b \\ a+b \end{pmatrix}$ . So

anything in the column span has the same first and last coordinate. So  $c = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$  is not in the column span.

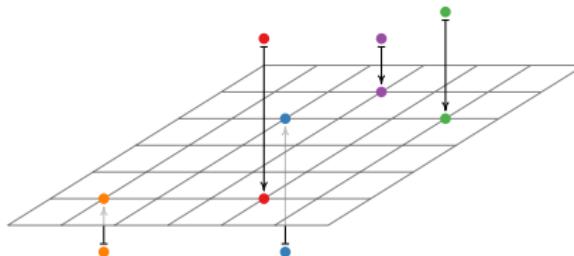
# Matrix Transformations

Geometric example

Let  $A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$  and let  $T(x) = Ax$ , so  $T: \mathbf{R}^3 \rightarrow \mathbf{R}^3$ . Then

$$T \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x \\ y \\ 0 \end{pmatrix}.$$

This is *projection onto the xy-axis*. Picture:



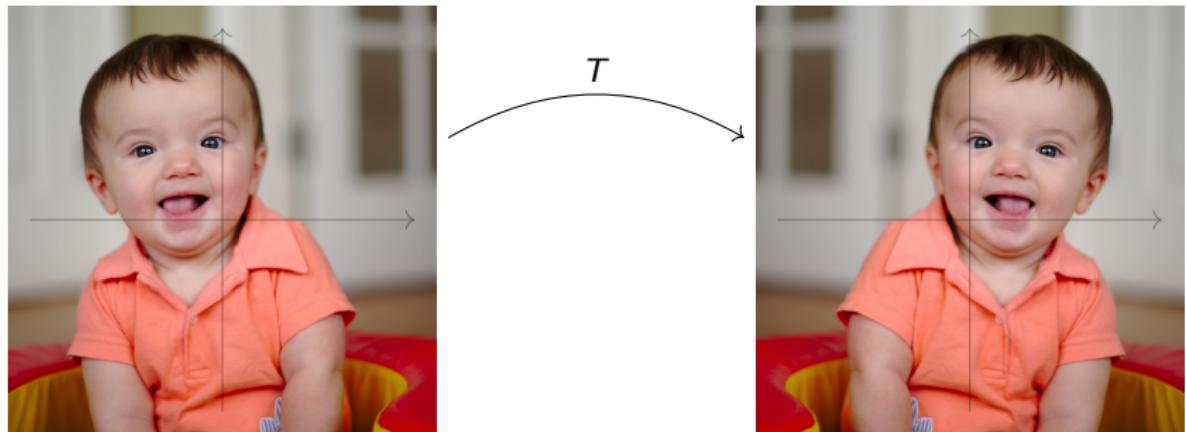
# Matrix Transformations

## Geometric example

Let  $A = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}$  and let  $T(x) = Ax$ , so  $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$ . Then

$$T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -x \\ y \end{pmatrix}.$$

This is *reflection over the y-axis*. Picture:

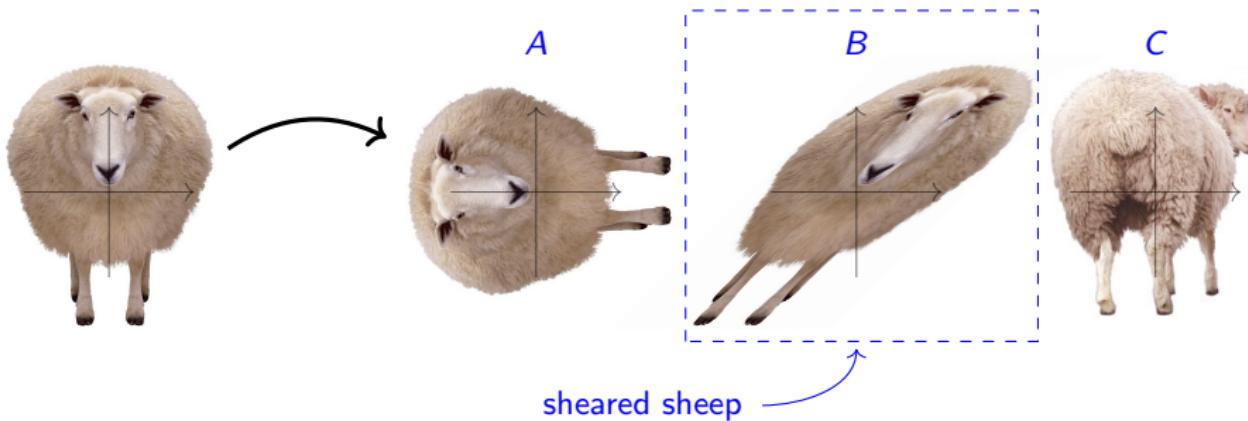


Let  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$  and let  $T(x) = Ax$ , so  $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$ . ( $T$  is called a **shear**.)

Poll

What does  $T$  do to this sheep?

Hint: first draw a picture what it does to the box *around* the sheep.



## Linear Transformations

Recall: If  $A$  is a matrix,  $u, v$  are vectors, and  $c$  is a scalar, then

$$A(u + v) = Au + Av \quad A(cv) = cAv.$$

So if  $T(x) = Ax$  is a matrix transformation then,

$$T(u + v) = T(u) + T(v) \quad T(cv) = cT(v).$$

This property is so special that it has its own name.

### Definition

A transformation  $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$  is **linear** if it satisfies the above [equations](#) for all vectors  $u, v$  in  $\mathbf{R}^n$  and all scalars  $c$ .

In other words,  $T$  “respects” addition and scalar multiplication.

Check: if  $T$  is linear, then

$$T(0) = 0 \quad T(cu + dv) = cT(u) + dT(v)$$

for all vectors  $u, v$  and scalars  $c, d$ . More generally,

$$T(c_1v_1 + c_2v_2 + \cdots + c_nv_n) = c_1T(v_1) + c_2T(v_2) + \cdots + c_nT(v_n).$$

In engineering this is called **superposition**.

# Linear Transformations

## Dilation

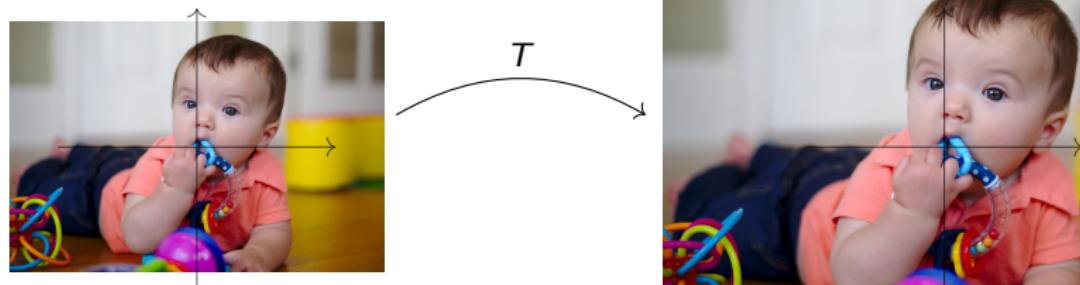
Define  $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$  by  $T(x) = 1.5x$ . Is  $T$  linear? Check:

$$T(u + v) = 1.5(u + v) = 1.5u + 1.5v = T(u) + T(v)$$

$$T(cv) = 1.5(cv) = c(1.5v) = c(Tv).$$

So  $T$  satisfies the two equations, hence  $T$  is linear.

This is called **dilation** or **scaling** (by a factor of 1.5). Picture:



# Linear Transformations

## Rotation

Define  $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$  by

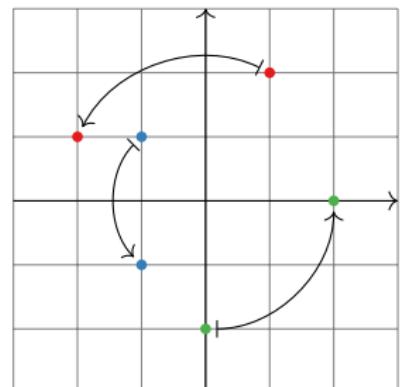
$$T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -y \\ x \end{pmatrix}.$$

Is  $T$  linear? Check:

$$T \left( \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} + \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} \right) = \begin{pmatrix} -u_2 \\ u_1 \end{pmatrix} + \begin{pmatrix} -v_2 \\ v_1 \end{pmatrix} = \begin{pmatrix} -(u_2 + v_2) \\ (u_1 + v_1) \end{pmatrix} = T \begin{pmatrix} u_1 + u_2 \\ v_1 + v_2 \end{pmatrix}$$

$$T \left( c \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} \right) = T \begin{pmatrix} cv_1 \\ cv_2 \end{pmatrix} = \begin{pmatrix} -cv_2 \\ cv_1 \end{pmatrix} = c \begin{pmatrix} -v_2 \\ v_1 \end{pmatrix} = cT \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}.$$

So  $T$  satisfies the two equations, hence  $T$  is linear. This is called **rotation** (by  $90^\circ$ ). Picture:



## Section 1.9

### The Matrix of a Linear Transformation

# Linear Transformations are Matrix Transformations

## Definition

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

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

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

$$A = \begin{pmatrix} | & | & | & | \\ T(e_1) & T(e_2) & \cdots & T(e_n) \\ | & | & | & | \end{pmatrix}.$$

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

In particular, *every linear transformation is a matrix transformation*.

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

## Linear Transformations are Matrix Transformations

Continued

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

$$\begin{aligned} T \begin{pmatrix} x \\ y \\ z \end{pmatrix} &= T \left( x \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + y \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + z \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \right) \\ &= T(xe_1 + ye_2 + ze_3) \\ &= xT(e_1) + yT(e_2) + zT(e_3) \\ &= \begin{pmatrix} | & | & | \\ T(e_1) & T(e_2) & T(e_3) \\ | & | & | \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} \\ &= A \begin{pmatrix} x \\ y \\ z \end{pmatrix}. \end{aligned}$$

So when we think of a matrix as a function from  $\mathbf{R}^n$  to  $\mathbf{R}^m$ , it's the same as thinking of a linear transformation.

## Linear Transformations are Matrix Transformations

### Example

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

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

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

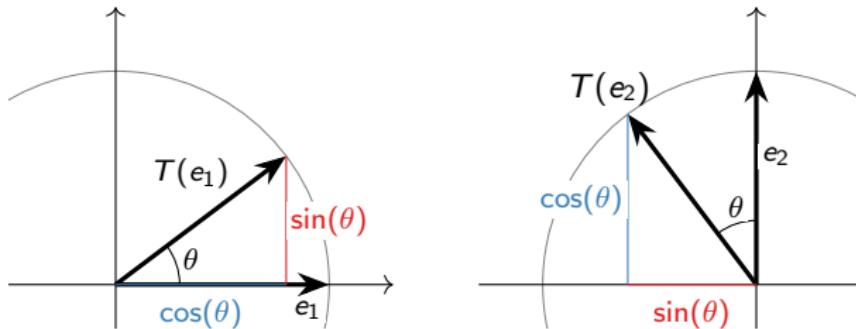
Example

## Question

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

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

(Check linearity...)



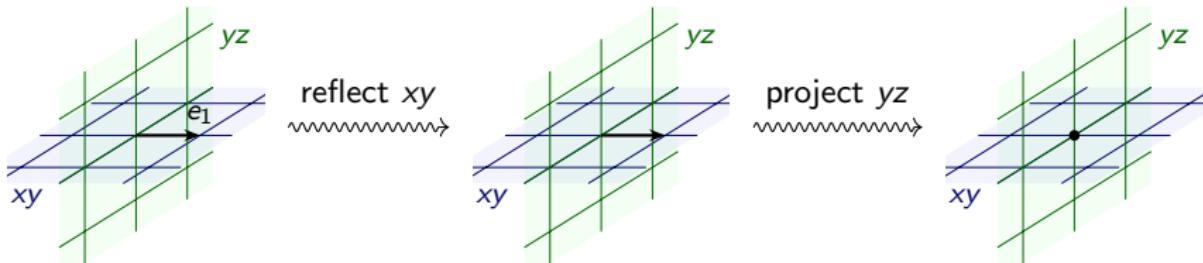
$$\left. \begin{array}{l} T(e_1) = \begin{pmatrix} \cos(\theta) \\ \sin(\theta) \end{pmatrix} \\ T(e_2) = \begin{pmatrix} -\sin(\theta) \\ \cos(\theta) \end{pmatrix} \end{array} \right\} \implies A = \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix} \quad \left(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

# 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?



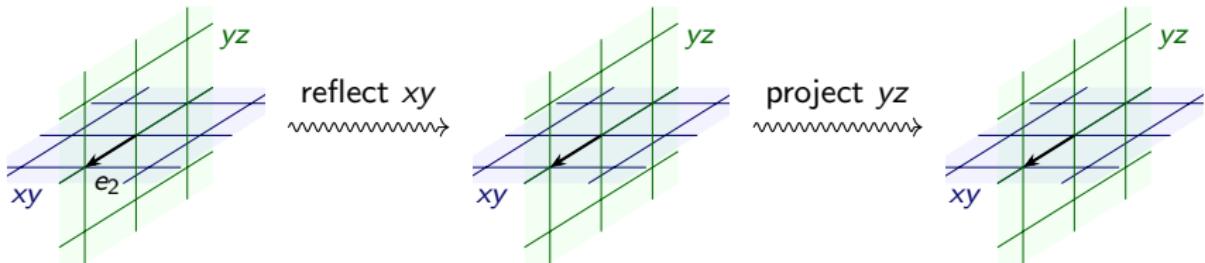
$$T(e_1) = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}.$$

# 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?



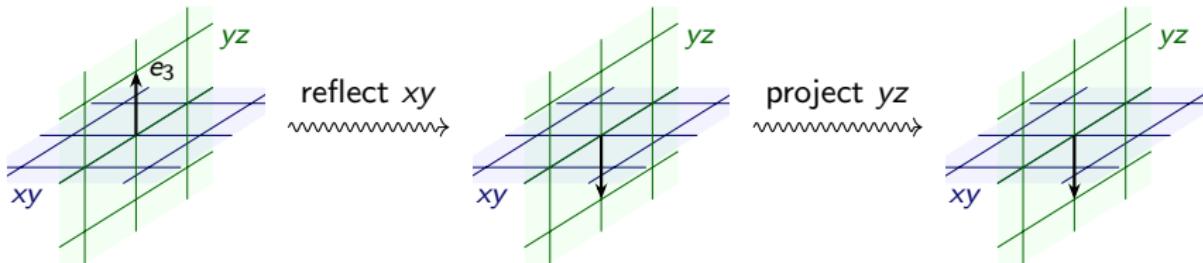
$$T(e_2) = e_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}.$$

# 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?

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

## Other Geometric Transformations

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.