

# Section 1.8

## Introduction to Linear Transformations

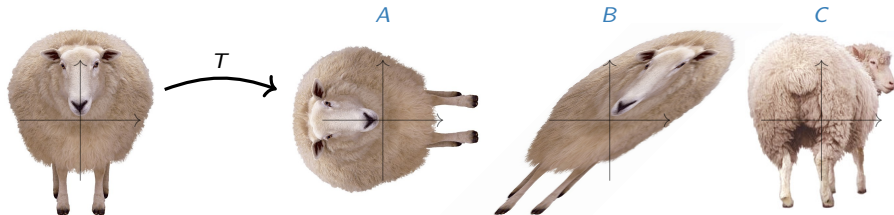
# Motivation

Let  $A$  be an  $m \times n$  matrix. For  $Ax = b$  we can describe

- ▶ the **solution set**: all  $x$  in  $\mathbf{R}^n$  making the **equation true**.
- ▶ the *column span*: the set of all  $b$  in  $\mathbf{R}^m$  making the *equation consistent*.

It turns out these two sets are *very closely related* to each other.

**Geometry matrices**: *linear transformation* from  $\mathbf{R}^n$  to  $\mathbf{R}^m$ .



# Transformations

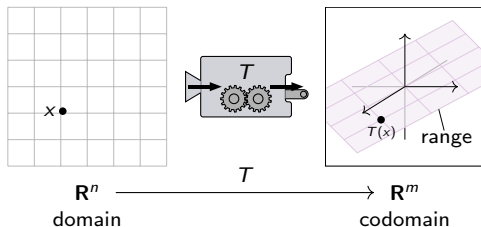
## Definition

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  $\mathbf{R}^m$ .

- ▶  $\mathbf{R}^n$  is called the **domain** of  $T$  (the inputs).
- ▶  $\mathbf{R}^m$  is called the **codomain** of  $T$  (the outputs).
- ▶ For  $x$  in  $\mathbf{R}^n$ , the vector  $T(x)$  in  $\mathbf{R}^m$  is the **image** of  $x$  under  $T$ .
- ▶ The set of all images  $\{T(x) \mid x \text{ in } \mathbf{R}^n\}$  is the **range** of  $T$ .

## Notation:

$T: \mathbf{R}^n \rightarrow \mathbf{R}^m$  means  $T$  is a transformation from  $\mathbf{R}^n$  to  $\mathbf{R}^m$ .



Think of  $T$  as a “*machine*”

- ▶ takes  $x$  as an input
- ▶ *gives you*  $T(x)$  as the output.

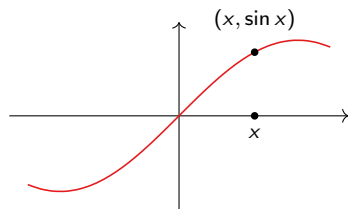
## Functions from Calculus

Many of the functions you know have domain and codomain  $\mathbf{R}$ .

$$\text{For example, } f: \mathbf{R} \longrightarrow \mathbf{R} \quad f(x) = x^2$$

Often times *we omit the name  $f(x)$  of the function* “ $x^2$ ”.

You may be used to thinking of a function in terms of its graph. E.g.,



The horizontal axis is the *domain*, and the vertical axis is the *codomain*.

This is fine when the domain and codomain are  $\mathbf{R}$ , but **it's hard to do when they're  $\mathbf{R}^2$  and  $\mathbf{R}^3$ !** You *need five dimensions* to draw that graph.

# Matrix Transformations

## 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  $A$** . It is *a span of vectors in the codomain*.

Your life will be much easier if you just remember these.

# 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  $T(v) = Av = b$ . We know how to do that:

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

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  $v$  in  $\mathbf{R}^2$  with  $T(v) = c$ ?

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

The solution set of  $Ax = c$  is a translate of the *solution set* of  $Ax = b$  (from before), which has *one vector in it*.

So the solution set to  $Ax = c$  has only one vector.

*So no!*

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

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

In other words, find  $c$  not in the column span of  $A$  (i.e., the range of  $T$ ).

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

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 (for example).

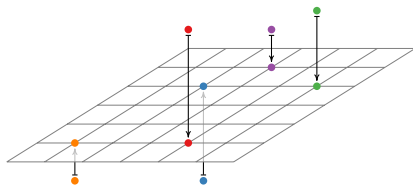
# Matrix Transformations

## Projection

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:





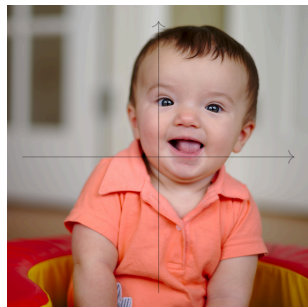
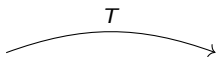
# Matrix Transformations

## Reflection

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:



# 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,

$$T(u+v) = T(u)+T(v) \quad \text{and} \quad T(cu) = cT(u)$$

## 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, (in engineering this is called **superposition**)

$$T(c_1 v_1 + c_2 v_2 + \cdots + c_n v_n) = c_1 T(v_1) + c_2 T(v_2) + \cdots + c_n T(v_n).$$

# Linear Transformations

## Dilation

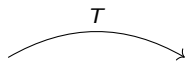
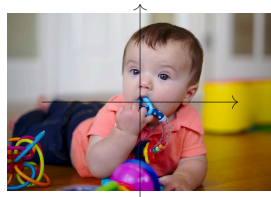
Define  $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$  by  $T(x) = 1.5x$ . Is  $T$  linear? Check:

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

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

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:

$$T \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} -2 \\ 1 \end{pmatrix}$$

$$T \begin{pmatrix} -1 \\ 1 \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \end{pmatrix}$$

$$T \begin{pmatrix} 0 \\ -2 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \end{pmatrix}$$

