

Section 1.8/1.9

Linear Transformations

Motivation

Let A be a matrix, and consider the matrix equation $b = Ax$. If we vary x , we can think of this as a *function* of x .

Many functions in real life—the *linear* transformations—come from matrices in this way.

It makes us happy when a function comes from a matrix, because then questions about the function translate into questions a matrix, which we can usually answer.

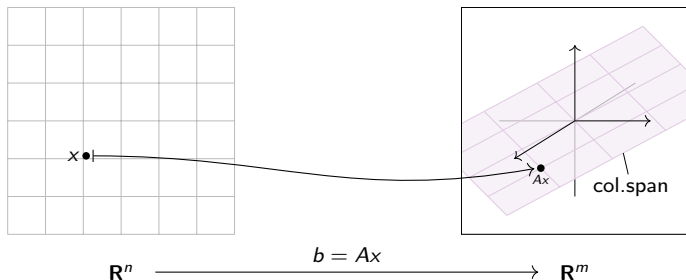
For this reason, we want to study matrices as functions.

Matrices as Functions

Change in Perspective. Let A be a matrix with m rows and n columns. Let's think about the matrix equation $b = Ax$ as a *function*.

- ▶ The independent variable (the input) is x , which is a vector in \mathbf{R}^n .
- ▶ The dependent variable (the output) is b , which is a vector in \mathbf{R}^m .

As you vary x , then $b = Ax$ also varies. The set of all possible output vectors b is the column span of A .



[interactive 1]

[interactive 2]

Matrices as Functions

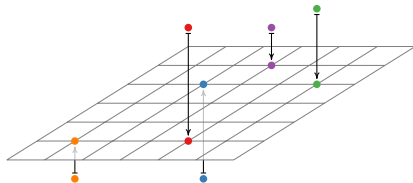
Projection

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

In the equation $Ax = b$, the input vector x is in \mathbf{R}^3 and the output vector b is in \mathbf{R}^3 . Then

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

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



[interactive]

Matrices as Functions

Reflection

$$A = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}$$

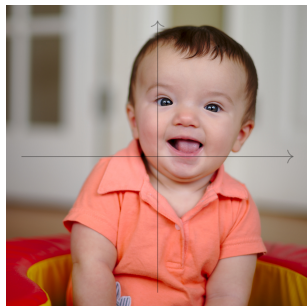
In the equation $Ax = b$, the input vector x is in \mathbf{R}^2 and the output vector b is in \mathbf{R}^2 . Then

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

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



$$b = Ax$$



[interactive]

Matrices as Functions

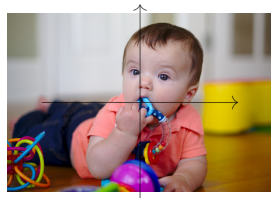
Dilation

$$A = \begin{pmatrix} 1.5 & 0 \\ 0 & 1.5 \end{pmatrix}$$

In the equation $Ax = b$, the input vector x is in \mathbf{R}^2 and the output vector b is in \mathbf{R}^2 .

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

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



$$b = Ax$$



[interactive]

Matrices as Functions

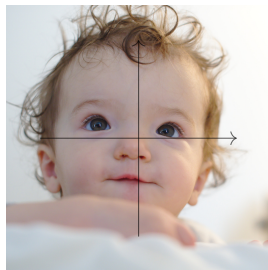
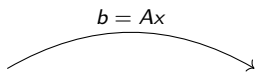
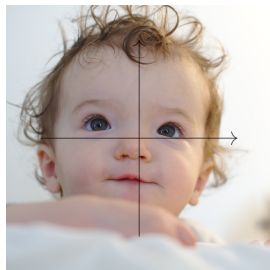
Identity

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

In the equation $Ax = b$, the input vector x is in \mathbf{R}^2 and the output vector b is in \mathbf{R}^2 .

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

This is *the identity transformation which does nothing*. Picture:



[interactive]

Matrices as Functions

Rotation

$$A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$$

In the equation $Ax = b$, the input vector x is in \mathbf{R}^2 and the output vector b is in \mathbf{R}^2 . Then

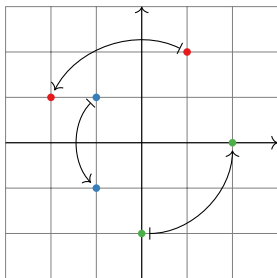
$$A \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -y \\ x \end{pmatrix}.$$

What is this? Let's plug in a few points and see what happens.

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

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

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



It looks like *counterclockwise rotation by 90°* .

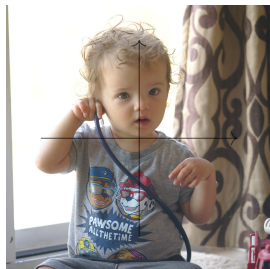
Matrices as Functions

Rotation

$$A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$$

In the equation $Ax = b$, the input vector x is in \mathbf{R}^2 and the output vector b is in \mathbf{R}^2 . Then

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



$$b = Ax$$



[interactive]

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

Transformations

Motivation

We have been drawing pictures of what it looks like to multiply a matrix by a vector, as a function of the vector.

Now let's go the other direction. Suppose we have a function, and we want to know, does it come from a matrix?

Example

For a vector x in \mathbf{R}^2 , let $T(x)$ be the counterclockwise rotation of x by an angle θ . Is $T(x) = Ax$ for some matrix A ?

If $\theta = 90^\circ$, then we know $T(x) = Ax$, where

$$A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}.$$

But for general θ , it's not clear.

Our next goal is to answer this kind of question.

Transformations

Vocabulary

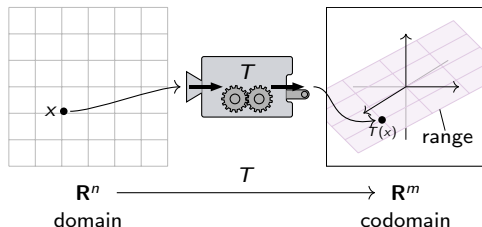
Definition

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 .
Notation: $x \mapsto T(x)$.
- ▶ The set of all images $\{T(x) \mid x \text{ in } \mathbf{R}^n\}$ is the **range** of T .

Notation:

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



It may help to think of T as a “machine” that takes x as an input, and gives you $T(x)$ as the output.

Functions from Calculus

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

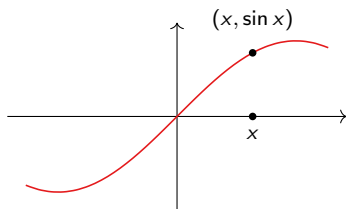
$$\sin: \mathbf{R} \rightarrow \mathbf{R} \quad \sin(x) = \left(\begin{array}{l} \text{the length of the opposite edge over the} \\ \text{hypotenuse of a right triangle with angle} \\ x \text{ in radians} \end{array} \right)$$

Note how I've written down the *rule* that defines the function \sin .

$$f: \mathbf{R} \rightarrow \mathbf{R} \quad f(x) = x^2$$

Note that " x^2 " is sloppy (but common) notation for a function: it doesn't have a name!

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



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 .

For example, if $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$ and $T(x) = Ax$ then

$$T \begin{pmatrix} -1 \\ -2 \\ -3 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \begin{pmatrix} -1 \\ -2 \\ -3 \end{pmatrix} = \begin{pmatrix} -14 \\ -32 \end{pmatrix}.$$

- ▶ 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 :

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

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.

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

We could draw a picture, or notice that if $c = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$, then our matrix equation translates into

$$x + y = 1 \quad y = 2 \quad x + y = 3,$$

which is obviously inconsistent.

Matrix Transformations

Non-Example

Note: All of these questions are questions about *the transformation* T ; it still makes sense to ask them in the absence of the matrix A .

The fact that T comes from a matrix means that these questions translate into questions about a matrix, which we know how to do.

Non-example: $T: \mathbf{R}^2 \rightarrow \mathbf{R}^3$ $T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} \sin x \\ xy \\ \cos y \end{pmatrix}$

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

Note the question still makes sense, although T has no hope of being a matrix transformation.

By the way,

$$T \begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} \sin 0 \\ 0 \cdot 0 \\ \cos 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} \sin \pi \\ 0 \cdot \pi \\ \cos 0 \end{pmatrix} = T \begin{pmatrix} \pi \\ 0 \end{pmatrix},$$

so the answer is yes.

Matrix Transformations

Picture

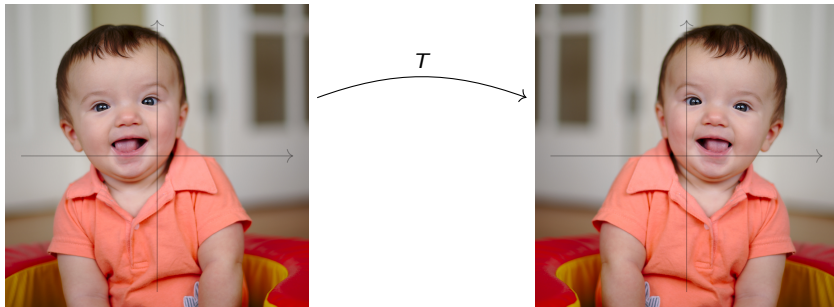
The picture of a matrix transformation is the same as the pictures we've been drawing all along. Only the language is different. Let

$$A = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} \quad \text{and let} \quad T(x) = Ax,$$

so $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$. Then

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

which is still is *reflection over the y-axis*. Picture:



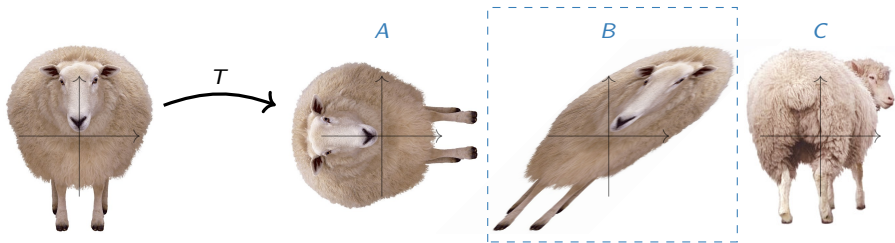
[Was not done in class]

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.



[interactive]

sheared sheep

So, which transformations actually come from matrices?

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,

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

Any matrix transformation has to satisfy this property. 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**.

Summary

- ▶ We can think of $b = Ax$ as a **transformation** with input x and output b . This gives us a way to draw pictures of the geometry of a matrix.
- ▶ There are lots of questions that one can ask about transformations.
- ▶ We like transformations that come from matrices, because questions about those transformations turn into questions about matrices.
- ▶ **Linear transformations** are the transformations that come from matrices.

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.

Note: T is a matrix transformation!

$$T(x) = \begin{pmatrix} 1.5 & 0 \\ 0 & 1.5 \end{pmatrix} x,$$

as we checked before.

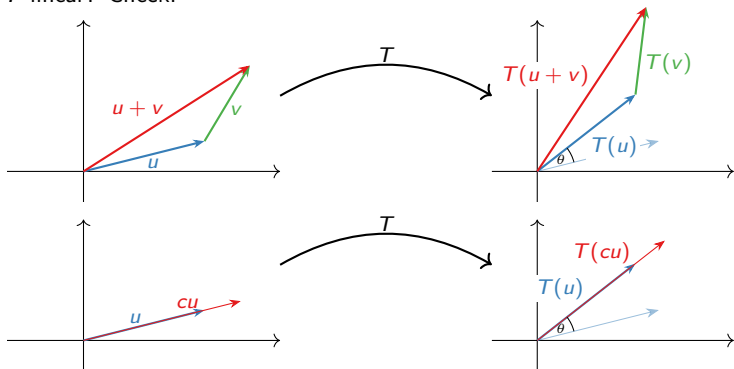
Linear Transformations

Rotation

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

$T(x)$ = the vector x rotated counterclockwise by an angle of θ .

Is T linear? Check:



The pictures show $T(u) + T(v) = T(u + v)$ and $T(cu) = cT(u)$.

Since T satisfies the two equations, T is linear.

Linear Transformations

Non-example

Is every transformation a linear transformation?

No! For instance, $T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} \sin x \\ xy \\ \cos y \end{pmatrix}$ is not linear.

Why? We have to check the two defining properties. Let's try the second:

$$T \left(c \begin{pmatrix} x \\ y \end{pmatrix} \right) = \begin{pmatrix} \sin(cx) \\ (cx)(cy) \\ \cos(cy) \end{pmatrix} \stackrel{?}{=} c \begin{pmatrix} \sin x \\ xy \\ \cos y \end{pmatrix} = cT \begin{pmatrix} x \\ y \end{pmatrix}$$

Not necessarily: if $c = 2$ and $x = \pi$, $y = \pi$, then

$$T \left(2 \begin{pmatrix} \pi \\ \pi \end{pmatrix} \right) = T \begin{pmatrix} 2\pi \\ 2\pi \end{pmatrix} = \begin{pmatrix} \sin 2\pi \\ 2\pi \cdot 2\pi \\ \cos 2\pi \end{pmatrix} = \begin{pmatrix} 0 \\ 4\pi^2 \\ 1 \end{pmatrix}$$

$$2T \begin{pmatrix} \pi \\ \pi \end{pmatrix} = 2 \begin{pmatrix} \sin \pi \\ \pi \cdot \pi \\ \cos \pi \end{pmatrix} = \begin{pmatrix} 0 \\ 2\pi^2 \\ -2 \end{pmatrix}.$$

So T fails the second property. **Conclusion:** T is *not* a matrix transformation!

(We could also have noted $T(0) \neq 0$.)

Poll

Which of the following transformations are linear?

$$\text{A. } T \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} |x_1| \\ x_2 \end{pmatrix} \quad \text{B. } T \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 2x_1 + x_2 \\ x_1 - 2x_2 \end{pmatrix}$$

$$\text{C. } T \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 x_2 \\ x_2 \end{pmatrix} \quad \text{D. } T \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 2x_1 + 1 \\ x_1 - 2x_2 \end{pmatrix}$$

A. $T \left(\begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} -1 \\ 0 \end{pmatrix} \right) = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \neq \begin{pmatrix} 2 \\ 0 \end{pmatrix} = T \begin{pmatrix} 1 \\ 0 \end{pmatrix} + T \begin{pmatrix} -1 \\ 0 \end{pmatrix}$, so not linear.

B. Linear.

C. $T \left(2 \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right) = \begin{pmatrix} 4 \\ 2 \end{pmatrix} \neq 2T \begin{pmatrix} 1 \\ 1 \end{pmatrix}$, so not linear.

D. $T \begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \neq 0$, so not linear.

Remark: in fact, T is linear if and only if each entry of the output is a linear function of the entries of the input, with no constant terms. Check this!

The Matrix of a Linear Transformation

We will see that a *linear* transformation T is a matrix transformation:

$$T(x) = Ax.$$

But what matrix does T come from? What is A ?

Here's how to compute it.

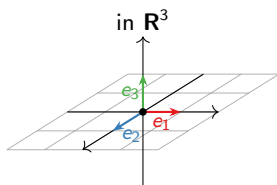
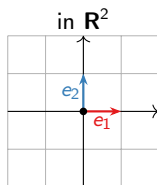
Unit Coordinate Vectors

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

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

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

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

Why is a linear transformation a matrix transformation?

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

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

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?

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

Check:

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

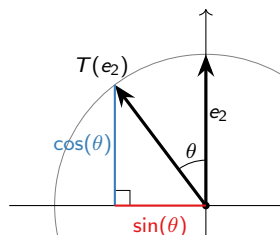
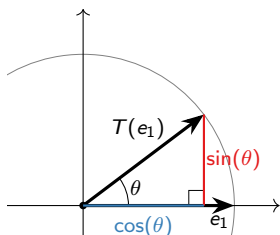
Linear Transformations are Matrix Transformations

Example

Question

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

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



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

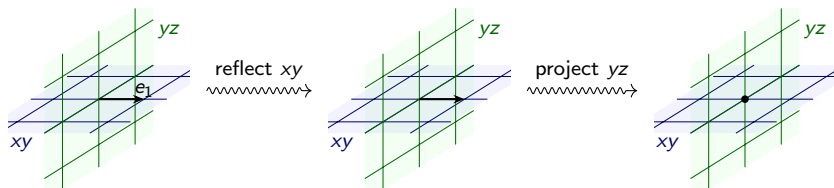
Linear Transformations are Matrix Transformations

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?

[interactive]



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

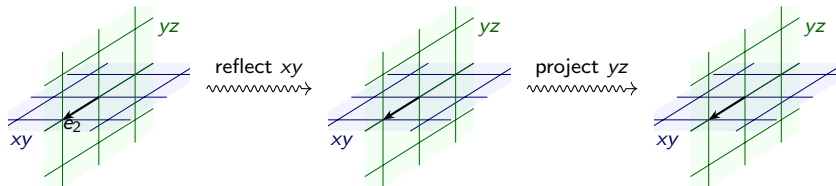
Linear Transformations are Matrix Transformations

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?

[interactive]



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

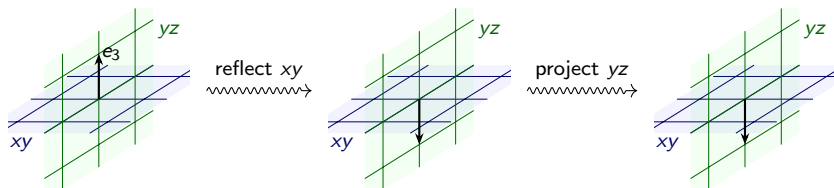
Linear Transformations are Matrix Transformations

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?

[interactive]



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

Linear Transformations are Matrix Transformations

Example, continued

Question

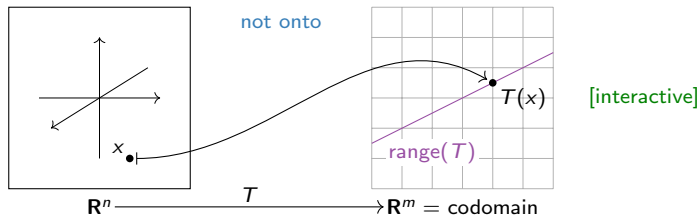
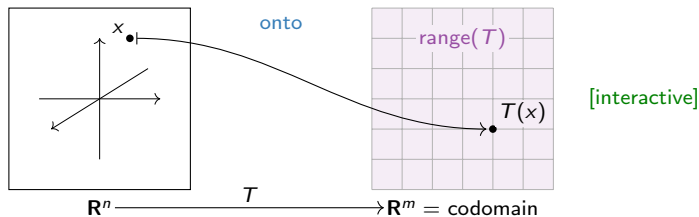
What is the matrix for the linear transformation $T: \mathbf{R}^3 \rightarrow \mathbf{R}^3$ that reflects through the xy -plane and then projects onto the yz -plane?

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

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, for every b in \mathbf{R}^m , the equation $T(x) = b$ has at least one solution. Or, 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: $m \leq n$.

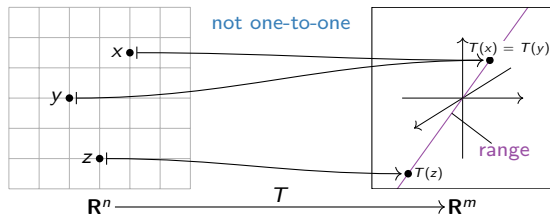
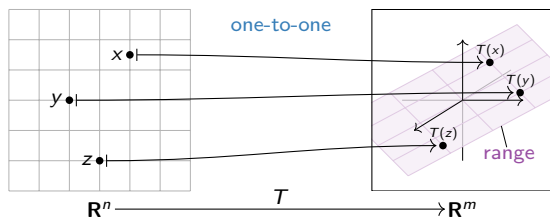
$$\begin{pmatrix} 1 & 0 & * & 0 & * \\ 0 & 1 & * & 0 & * \\ 0 & 0 & 0 & 1 & * \end{pmatrix}$$

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, for every b in \mathbf{R}^m , the equation $T(x) = b$ has *at most one* solution x . Or, different inputs have different outputs. Note that *not* one-to-one means at least two 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*: $n \leq m$.

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

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

Summary

- ▶ **Linear transformations** are the transformations that come from matrices.
- ▶ The **unit coordinate vectors** e_1, e_2, \dots are the unit vectors in the positive direction along the coordinate axes.
- ▶ You compute the columns of the matrix for a linear transformation by plugging in the unit coordinate vectors.
- ▶ A transformation T is **one-to-one** if $T(x) = b$ has *at most one* solution, for every b in \mathbf{R}^m .
- ▶ A transformation T is **onto** if $T(x) = b$ has *at least one* solution, for every b in \mathbf{R}^m .
- ▶ Two of the most basic questions one can ask about a transformation is whether it is one-to-one or onto.
- ▶ There are lots of equivalent conditions for a linear transformation to be one-to-one and/or onto, in terms of its matrix.