

Announcements

Wednesday, October 18

- ▶ The second midterm is on **this Friday, October 20**.
 - ▶ The exam covers §§1.7, 1.8, 1.9, 2.1, 2.2, 2.3, 2.8, and 2.9.
 - ▶ About half the problems will be conceptual, and the other half computational.
 - ▶ Note that this midterm covers **more material** than the first!
- ▶ There is a practice midterm posted on the website. It is identical in format to the real midterm (although there may be $\pm 1-2$ problems).
- ▶ Study tips:
 - ▶ There are lots of problems at the end of each section in the book, and at the end of the chapter, for practice.
 - ▶ Make sure to **learn the theorems** and **learn the definitions**, and understand what they mean. There is a reference sheet on the website.
 - ▶ Sit down to do the practice midterm in 50 minutes, with no notes.
 - ▶ Come to office hours!
- ▶ WeBWork 2.8, 2.9 are due today at 11:59pm.
- ▶ **Double Rabinoffice hours this week:** Monday, 1–3pm; Tuesday, 9–11am; Thursday, 9–11am; Thursday, 12–2pm.
- ▶ **TA review session:** Today, 7:15–9pm, Culc 144.

Midterm 2

Review Slides

Transformations

Vocabulary

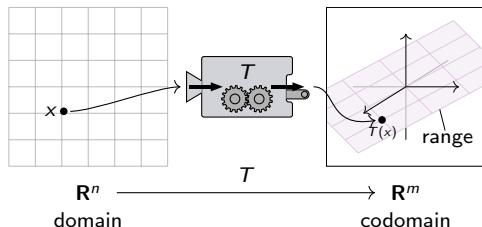
Definition

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.

Matrix Transformations

If A is an $m \times n$ matrix, then

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

is a **matrix transformation**.

These are the kinds of transformations we can use linear algebra to study, because they come from *matrices*.

Example: $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$

$$T \begin{pmatrix} x \\ y \\ z \end{pmatrix} =$$

(Note we've written a *formula* for T that doesn't a priori have anything to do with matrices.)

Questions about Transformations

Here are some natural questions that one can ask about a general transformation (not just on the midterm, but in the real world too):

Question: What kind of vectors can you input into T ? What kind of vectors do you get out? In other words, what are the domain and codomain?

Answer for $T(x) = Ax$: Inputs are in \mathbf{R}^n , where n is the number of *columns* of T . Outputs are in \mathbf{R}^m , where m is the number of *rows* of A . (Cf. previous slide.)

Question: For which b does $T(x) = b$ have a solution? In other words, what is the range of T ?

Answer for $T(x) = Ax$: The range is $\text{Col } A$, the span of the columns: $T(x) = Ax$ is a linear combination of the columns of A .

Question: Is T one-to-one, onto, and/or invertible?

Answer for $T(x) = Ax$: on the next slides

One-to-one and onto

Definition

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

- ▶ **one-to-one** if $T(x) = b$ has at *most* one solution for every b in \mathbf{R}^m
- ▶ **onto** if $T(x) = b$ has at *least* one solution for every b in \mathbf{R}^m

Picture: [\[interactive\]](#)

This is neither one-to-one nor onto.

- ▶ Can you find two different solutions to $T(x) = 0$?
- ▶ Can you find a b such that $T(x) = b$ has no solution?

Picture: [\[interactive\]](#)

This is onto but not one-to-one.

- ▶ Can you find two different solutions to $T(x) = 0$?

Picture: [\[interactive\]](#)

This is one-to-one and onto.

One-to-one and Onto Matrix Transformations

Theorem

Let $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$ be a matrix 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 *column*.

Theorem

Let $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$ be a matrix 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*

Linear Transformations

Question: How do you know if a transformation is a matrix transformation or not?

Definition

A transformation $T: \mathbf{R}^n \rightarrow \mathbf{R}^m$ is **linear** if it satisfies the the equations

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

for all vectors u, v in \mathbf{R}^n and all scalars c . ($\implies T(0) = 0$)

Theorem

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

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

So a linear transformation is a matrix transformation, where you haven't computed the matrix yet.

Important

You compute the columns of the matrix for T by plugging in e_1, e_2, \dots, e_n .

Examples

Example: $T: \mathbf{R} \rightarrow \mathbf{R}$ defined by $T(x) = x + 1$.

This is not linear: $T(0) = 1 \neq 0$.

Example: $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$ defined by rotation by θ degrees. Is T linear? Check:

The pictures show $T(u) + T(v) = T(u + v)$ and $T(cu) = cT(u)$, so T is linear.

Examples

Continued

Example: $T: \mathbf{R}^2 \rightarrow \mathbf{R}^2$ defined by rotation by θ degrees. What is the standard matrix?

Examples

Continued

Example: $T: \mathbb{R}^3 \rightarrow \mathbb{R}^2$ defined by

$$T \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2x + 3y - z \\ y + z \end{pmatrix}.$$

Is T linear? Check $T(u + v) = T(u) + T(v)$:

Note we're treating u and v as *unknown* vectors: this has to work for all vectors u and v !

Examples

Continued

Example: $T: \mathbb{R}^3 \rightarrow \mathbb{R}^2$ defined by

$$T \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2x + 3y - z \\ y + z \end{pmatrix}.$$

Is T linear? Check $T(cu) = cT(u)$:

Conclusion: T is linear.

Examples

Continued

Example: $T: \mathbf{R}^3 \rightarrow \mathbf{R}^2$ defined by

$$T \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2x + 3y - z \\ y + z \end{pmatrix}.$$

We know it is linear, so it is a matrix transformation. What is its standard matrix A ?

Subspaces

Definition

A **subspace** of \mathbf{R}^n is a subset V of \mathbf{R}^n satisfying:

1. The zero vector is in V . "not empty"
2. If u and v are in V , then $u + v$ is also in V . "closed under addition"
3. If u is in V and c is in \mathbf{R} , then cu is in V . "closed under \times scalars"

A subspace is a span, and a span is a subspace.

Important examples of subspaces:

- ▶ The span of any set of vectors.
- ▶ The column space of a matrix.
- ▶ The null space of a matrix.
- ▶ The solution set of a system of homogeneous equations.
- ▶ All of \mathbf{R}^n and the zero subspace $\{0\}$.

Subspaces

What is the point?

The point of a subspace is to talk about a span without figuring out which vectors it's the span of.

Example: $A = \begin{pmatrix} 2 & 7 & -4 & 3 \\ 0 & 0 & 12 & 1 \\ 0 & 0 & 0 & -78 \end{pmatrix} \quad V = \text{Nul } A$

There are 3 pivots, so $\text{rank } A = 3$.

By the rank theorem, $\dim \text{Nul } A = 1$.

We know the null space is a line, but we never had to compute a spanning vector!