

## Announcements

Monday, September 10

- ▶ WeBWorK due on Wednesday at 11:59pm.
- ▶ The quiz on Friday covers §3.1 and §3.2.
- ▶ The first midterm is on **Friday, September 21**.
  - ▶ That is one week from this Friday.
  - ▶ Midterms happen during recitation.
  - ▶ The exam covers *through* §3.4.
- ▶ My office is Skiles 244 and Rabin office hours are: Mondays, 12–1pm; Wednesdays, 1–3pm.

## Section 3.3

### Matrix Equations

## Matrix $\times$ Vector

the first number is  
the number of rows

the second number is  
the number of columns

Let  $A$  be an  $m \times n$  matrix

$$A = \begin{pmatrix} | & | & & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & & | \end{pmatrix} \quad \text{with columns } v_1, v_2, \dots, v_n$$

### Definition

The **product** of  $A$  with a vector  $x$  in  $\mathbb{R}^n$  is the linear combination

$$Ax = \begin{pmatrix} | & | & & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & & | \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \stackrel{\text{def}}{=} x_1 v_1 + x_2 v_2 + \cdots + x_n v_n.$$

these must be equal

this means the equality  
is a *definition*

The output is a vector in  $\mathbb{R}^m$ .

Note that the number of **columns** of  $A$  has to equal the number of **rows** of  $x$ .

### Example

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

# Matrix Equations

An example

## Question

Let  $v_1, v_2, v_3$  be vectors in  $\mathbf{R}^3$ . How can you write the vector equation

$$2v_1 + 3v_2 - 4v_3 = \begin{pmatrix} 7 \\ 2 \\ 1 \end{pmatrix}$$

in terms of matrix multiplication?

## Matrix Equations

In general

Let  $v_1, v_2, \dots, v_n$ , and  $b$  be vectors in  $\mathbb{R}^m$ . Consider the vector equation

$$x_1 v_1 + x_2 v_2 + \cdots + x_n v_n = b.$$

It is equivalent to the **matrix equation**

$$Ax = b$$

where

$$A = \begin{pmatrix} | & | & \cdots & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & \cdots & | \end{pmatrix} \quad \text{and} \quad x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}.$$

Conversely, if  $A$  is any  $m \times n$  matrix, then

$$Ax = b \quad \begin{matrix} \text{is equivalent to the} \\ \text{vector equation} \end{matrix} \quad x_1 v_1 + x_2 v_2 + \cdots + x_n v_n = b$$

where  $v_1, \dots, v_n$  are the columns of  $A$ , and  $x_1, \dots, x_n$  are the entries of  $x$ .

## Linear Systems, Vector Equations, Matrix Equations, ...

We now have *four* equivalent ways of writing (and thinking about) linear systems:

1. As a system of equations:

$$\begin{aligned}2x_1 + 3x_2 &= 7 \\x_1 - x_2 &= 5\end{aligned}$$

2. As an augmented matrix:

$$\left( \begin{array}{cc|c} 2 & 3 & 7 \\ 1 & -1 & 5 \end{array} \right)$$

3. As a vector equation ( $x_1 v_1 + \cdots + x_n v_n = b$ ):

$$x_1 \begin{pmatrix} 2 \\ 1 \end{pmatrix} + x_2 \begin{pmatrix} 3 \\ -1 \end{pmatrix} = \begin{pmatrix} 7 \\ 5 \end{pmatrix}$$

4. As a matrix equation ( $Ax = b$ ):

$$\begin{pmatrix} 2 & 3 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 7 \\ 5 \end{pmatrix}$$

We will move back and forth freely between these over and over again, for the rest of the semester. Get comfortable with them now!

In particular, *all four have the same solution set*.

## Matrix $\times$ Vector

Another way

### Definition

A **row vector** is a matrix with one row. The product of a row vector of length  $n$  and a (column) vector of length  $n$  is

$$(a_1 \quad \cdots \quad a_n) \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \stackrel{\text{def}}{=} a_1 x_1 + \cdots + a_n x_n.$$

This is a scalar.

If  $A$  is an  $m \times n$  matrix with rows  $r_1, r_2, \dots, r_m$ , and  $x$  is a vector in  $\mathbf{R}^n$ , then

$$Ax = \begin{pmatrix} -r_1- \\ -r_2- \\ \vdots \\ -r_m- \end{pmatrix} x = \begin{pmatrix} r_1 x \\ r_2 x \\ \vdots \\ r_m x \end{pmatrix}$$

This is a vector in  $\mathbf{R}^m$  (again).

## Matrix $\times$ Vector

Both ways

### Example

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

Note this is the same as before:

Now you have *two* ways of computing  $Ax$ .

In the second, you calculate  $Ax$  one entry at a time.

The second way is usually the most convenient, but we'll use both.

In engineering, the first way corresponds to “superposition of states”, and the second is “taking a measurement”.

## Spans and Solutions to Equations

Let  $A$  be a matrix with columns  $v_1, v_2, \dots, v_n$ :

$$A = \begin{pmatrix} | & | & & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & & | \end{pmatrix}$$

Very Important Fact That Will Appear on Every Midterm and the Final

$Ax = b$  has a solution

$$\iff \text{there exist } x_1, \dots, x_n \text{ such that } A \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = b$$

"if and only if"

$$\begin{aligned} &\iff \text{there exist } x_1, \dots, x_n \text{ such that } x_1 v_1 + \cdots + x_n v_n = b \\ &\iff b \text{ is a linear combination of } v_1, \dots, v_n \\ &\iff b \text{ is in the span of the columns of } A. \end{aligned}$$

The last condition is geometric.

# Spans and Solutions to Equations

Example

## Question

Let  $A = \begin{pmatrix} 2 & 1 \\ -1 & 0 \\ 1 & -1 \end{pmatrix}$ . Does the equation  $Ax = \begin{pmatrix} 0 \\ 2 \\ 2 \end{pmatrix}$  have a solution?

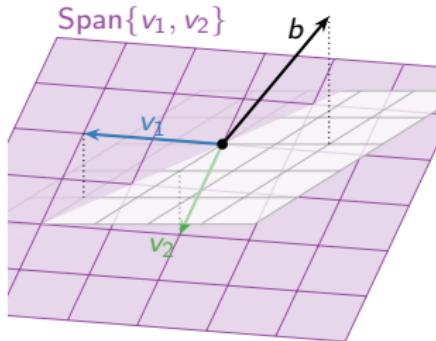
[interactive]

Columns of  $A$ :

$$v_1 = \begin{pmatrix} 2 \\ -1 \\ 1 \end{pmatrix} \quad v_2 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

Target vector:

$$b = \begin{pmatrix} 0 \\ 2 \\ 2 \end{pmatrix}$$



Is  $b$  contained in the span of the columns of  $A$ ? It sure doesn't look like it.

**Conclusion:**  $Ax = b$  is *inconsistent*.

## Spans and Solutions to Equations

Example, continued

### Question

Let  $A = \begin{pmatrix} 2 & 1 \\ -1 & 0 \\ 1 & -1 \end{pmatrix}$ . Does the equation  $Ax = \begin{pmatrix} 0 \\ 2 \\ 2 \end{pmatrix}$  have a solution?

Answer: Let's check by solving the matrix equation using row reduction.

In other words, the matrix equation

$$\begin{pmatrix} 2 & 1 \\ -1 & 0 \\ 1 & -1 \end{pmatrix} x = \begin{pmatrix} 0 \\ 2 \\ 2 \end{pmatrix}$$

has no solution, as the picture shows.

# Spans and Solutions to Equations

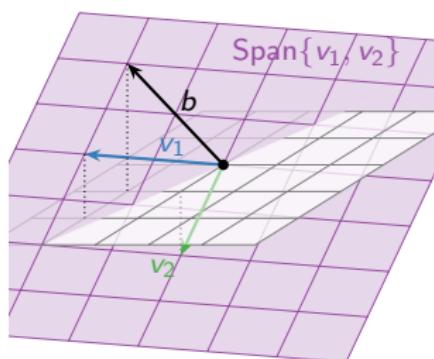
Example

## Question

Let  $A = \begin{pmatrix} 2 & 1 \\ -1 & 0 \\ 1 & -1 \end{pmatrix}$ . Does the equation  $Ax = \begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix}$  have a solution?

[interactive]

Columns of  $A$ :



$$v_1 = \begin{pmatrix} 2 \\ -1 \\ 1 \end{pmatrix} \quad v_2 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

Target vector:

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

Is  $b$  contained in the span of the columns of  $A$ ? It looks like it: in fact,

$$b = 1v_1 + (-1)v_2 \implies x = \begin{pmatrix} 1 \\ -1 \end{pmatrix}.$$

## Spans and Solutions to Equations

Example, continued

### Question

Let  $A = \begin{pmatrix} 2 & 1 \\ -1 & 0 \\ 1 & -1 \end{pmatrix}$ . Does the equation  $Ax = \begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix}$  have a solution?

Answer: Let's do this systematically using row reduction.

This is consistent with the picture on the previous slide:

$$1 \begin{pmatrix} 2 \\ -1 \\ 1 \end{pmatrix} - 1 \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix} \quad \text{or} \quad A \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix}.$$

Poll

## When Solutions Always Exist

Here are criteria for a linear system to *always* have a solution.

### Theorem

Let  $A$  be an  $m \times n$  (non-augmented) matrix. The following are equivalent:

1.  $Ax = b$  has a solution *for all*  $b$  in  $\mathbf{R}^m$ .
2. The span of the columns of  $A$  is all of  $\mathbf{R}^m$ .
3.  $A$  has a pivot in each row.

↑  
recall that this means  
that for given  $A$ , either they're  
all true, or they're all false

Why is (1) the same as (2)? This was the **Very Important** box from before.

Why is (1) the same as (3)? If  $A$  has a pivot in each row then its reduced row echelon form looks like this:

$$\left( \begin{array}{ccccc} 1 & 0 & * & 0 & * \\ 0 & 1 & * & 0 & * \\ 0 & 0 & 0 & 1 & * \end{array} \right) \quad \text{and } (A | b) \text{ reduces to this:} \quad \left( \begin{array}{ccccc|c} 1 & 0 & * & 0 & * & * \\ 0 & 1 & * & 0 & * & * \\ 0 & 0 & 0 & 1 & * & * \end{array} \right).$$

There's no  $b$  that makes it inconsistent, so there's always a solution. If  $A$  doesn't have a pivot in each row, then its reduced form looks like this:

$$\left( \begin{array}{ccccc} 1 & 0 & * & 0 & * \\ 0 & 1 & * & 0 & * \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \quad \text{and this can be made inconsistent:} \quad \left( \begin{array}{ccccc|c} 1 & 0 & * & 0 & * & 0 \\ 0 & 1 & * & 0 & * & 0 \\ 0 & 0 & 0 & 0 & 0 & 16 \end{array} \right).$$

# When Solutions Always Exist

Continued

## Theorem

Let  $A$  be an  $m \times n$  (non-augmented) matrix. The following are equivalent:

1.  $Ax = b$  has a solution *for all*  $b$  in  $\mathbf{R}^m$ .
2. The span of the columns of  $A$  is all of  $\mathbf{R}^m$ .
3.  $A$  has a pivot in each row.

In the following demos, the **violet** region is the span of the columns of  $A$ . This is the same as the set of all  $b$  such that  $Ax = b$  has a solution.

[example where the criteria are satisfied]

[example where the criteria are not satisfied]

## Properties of the Matrix–Vector Product

Let  $c$  be a scalar,  $u, v$  be vectors, and  $A$  a matrix.

- ▶  $A(u + v) = Au + Av$
- ▶  $A(cv) = cAv$

**Consequence:** If  $u$  and  $v$  are solutions to  $Ax = 0$ , then so is every vector in  $\text{Span}\{u, v\}$ . Why?

Important

The set of solutions to  $Ax = 0$  is a span.

## Summary

- ▶ We have four equivalent ways of writing a system of linear equations:
  1. As a system of equations.
  2. As an augmented matrix.
  3. As a vector equation.
  4. As a matrix equation  $Ax = b$ .
- ▶  $Ax = b$  is consistent if and only if  $b$  is in the span of the columns of  $A$ . The latter condition is geometric: you can draw pictures of it.
- ▶  $Ax = b$  is consistent for all  $b$  in  $\mathbb{R}^m$  if and only if the columns of  $A$  span  $\mathbb{R}^m$ .