

# Section 1.3-1.4

Vector and Matrix Equations

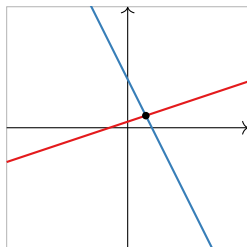
# Motivation

Linear algebra's *two viewpoints*:

- ▶ **Algebra**: systems of equations and their solution sets
- ▶ **Geometry**: intersections of points, lines, planes, etc.

$$\begin{aligned}x - 3y &= -3 \\ 2x + y &= 8\end{aligned}$$

↔



The **geometry** will give us *better insight into the properties* of systems of equations and their solution sets.

## Most Important Today: Spans and Solutions to Equations

Let  $\mathbf{b} \in \mathbf{R}^n$  and  $A$  be a matrix with columns  $v_1, v_2, \dots, v_n \in \mathbf{R}^n$ :

$$A = \left( \begin{array}{c|c|c|c} | & | & & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & & | \end{array} \right)$$

Very Important Fact

$Ax = b$  has a solution

$\iff b$  is a linear combination of  $v_1, \dots, v_n$

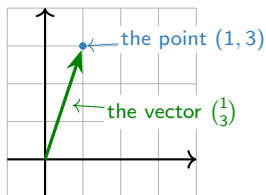
$\iff b$  is in the span of the columns of  $A$ .

The last condition is geometric.

# Vectors

Elements of  $\mathbf{R}^n$  can be considered *points*...

or **vectors**:  
arrows with a given  
*length and direction*.

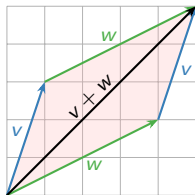


It is *convenient* to express **vectors** in  $\mathbf{R}^n$  as **matrices** with  $n$  rows and *one column*:

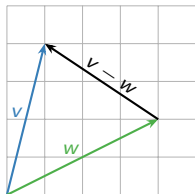
$$v = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$$

## Addition (*parallelogram law*) and Subtraction

**Addition:** Is the one that commutes



**Subtraction:** If you add  $\mathbf{v} - \mathbf{w}$  to  $\mathbf{w}$ , you get  $\mathbf{v}$ .



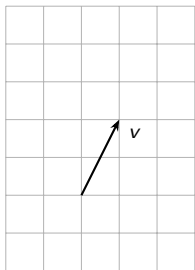
## Towards “linear spaces”

### Scalar multiples of a vector:

have the same *direction* but a different *length*.

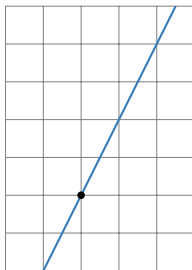
The scalar multiples of  $\mathbf{v}$  *form a line*.

Some multiples of  $\mathbf{v}$ .



$$\mathbf{v} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$
$$2\mathbf{v} = \begin{pmatrix} 2 \\ 4 \end{pmatrix}$$
$$-\frac{1}{2}\mathbf{v} = \begin{pmatrix} -\frac{1}{2} \\ -1 \end{pmatrix}$$
$$0\mathbf{v} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

All multiples of  $\mathbf{v}$ .



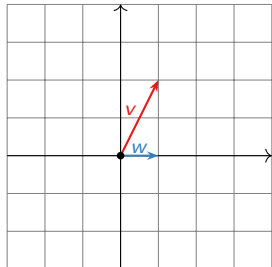
## Linear Combinations: *generate new vectors*

### Definition

$$\mathbf{w} = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \cdots + c_p \mathbf{v}_p$$

We call  $\mathbf{w}$  a **linear combination** of the vectors  $v_1, v_2, \dots, v_p$ , and the scalars  $c_1, c_2, \dots, c_p$  are called the *weights* or *coefficients*.

### Example



Let  $v = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$  and  $w = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ .

What are some linear combinations of  $v$  and  $w$ ?

- ▶  $v + w$
- ▶  $v - w$
- ▶  $2v + 0w$
- ▶  $2w$
- ▶  $-v$

# Matrix $\times$ Vector

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

$$A = \left( \begin{array}{c|c|c|c} | & | & \cdots & | \\ \mathbf{v}_1 & \mathbf{v}_2 & & \mathbf{v}_n \\ | & | & & | \end{array} \right) \quad \text{with columns } \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$$

## Definition

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

$$Ax = \left( \begin{array}{c|c|c|c} | & | & \cdots & | \\ \mathbf{v}_1 & \mathbf{v}_2 & & \mathbf{v}_n \\ | & | & & | \end{array} \right) \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \stackrel{\text{def}}{=} x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + \cdots + x_n \mathbf{v}_n.$$

- ▶ **Necessary:** Number of **columns** of  $A$  equals number of **rows** of  $x$ .
- ▶ The output is a vector in  $\mathbf{R}^m$ .



# Matrix $\times$ Vector

Another way

A **row vector** is a matrix with one row.

Dot product

The **product** of a row vector of length  $n$  and a (column) vector of length  $n$

$$(a_1 \quad \cdots \quad a_n) \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \stackrel{\text{def}}{=} a_1x_1 + \cdots + a_nx_n.$$

**is a scalar!**

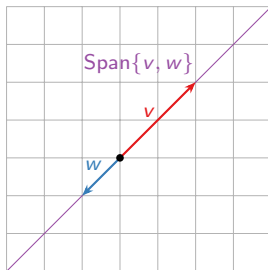
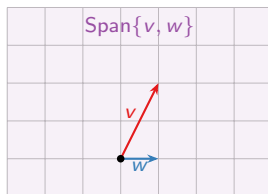
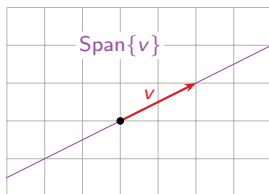
If  $A$  is an  $m \times n$  matrix with rows  $r_1, r_2, \dots, r_m$ , and  $x$  is a vector in  $\mathbf{R}^n$ , then

$$Ax = \begin{pmatrix} -r_1- \\ -r_2- \\ \vdots \\ -r_m- \end{pmatrix} x = \begin{pmatrix} r_1x \\ r_2x \\ \vdots \\ r_mx \end{pmatrix}$$

This is **a vector in  $\mathbf{R}^m$** .

## Pictures of Span in $R^2$

Drawing a picture of  $\text{Span}\{v_1, v_2, \dots, v_p\}$  is the same as drawing a picture of all linear combinations of  $v_1, v_2, \dots, v_p$ .



## Definition

Let  $v_1, v_2, \dots, v_p$  be vectors in  $\mathbf{R}^n$ . The **span** of  $v_1, v_2, \dots, v_p$  is the collection of **all linear combinations** of  $v_1, v_2, \dots, v_p$ , and is denoted  $\text{Span}\{v_1, v_2, \dots, v_p\}$ . In symbols:

**In other words:**

- ▶  $\text{Span}\{v_1, v_2, \dots, v_p\}$  is the subset **spanned/generated by**  $v_1, v_2, \dots, v_p$ .
- ▶ it's exactly the *collection of all  $b$  in  $\mathbf{R}^n$*  such that the *vector equation*

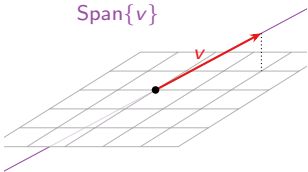
$$x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + \dots + x_p\mathbf{v}_p = \mathbf{b}$$

**is consistent** i.e., has a solution (unknowns  $x_1, x_2, \dots, x_p$ ).

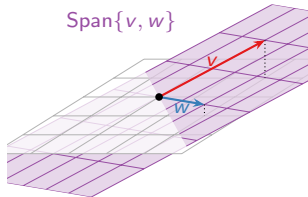


# Pictures of Span in $\mathbf{R}^3$

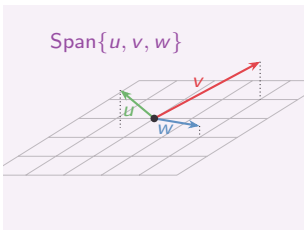
Span $\{v\}$



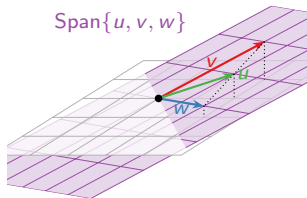
Span $\{v, w\}$



Span $\{u, v, w\}$



Span $\{u, v, w\}$



## Systems of Linear Equations

Is  $\begin{pmatrix} 8 \\ 16 \\ 3 \end{pmatrix}$  a linear combination of  $\begin{pmatrix} 1 \\ 2 \\ 6 \end{pmatrix}$  and  $\begin{pmatrix} -1 \\ -2 \\ -1 \end{pmatrix}$ ?

$$\begin{aligned}x - y &= 8 \\2x - 2y &= 16 \\6x - y &= 3\end{aligned}$$

matrix form  
~~~~~>

$$\left( \begin{array}{cc|c} 1 & -1 & 8 \\ 2 & -2 & 16 \\ 6 & -1 & 3 \end{array} \right)$$

row reduce  
~~~~~>

$$\left( \begin{array}{cc|c} 1 & 0 & -1 \\ 0 & 1 & -9 \\ 0 & 0 & 0 \end{array} \right)$$

solution  
~~~~~>

$$\begin{aligned}x &= -1 \\y &= -9\end{aligned}$$

Conclusion:

$$-\begin{pmatrix} 1 \\ 2 \\ 6 \end{pmatrix} - 9\begin{pmatrix} -1 \\ -2 \\ -1 \end{pmatrix} = \begin{pmatrix} 8 \\ 16 \\ 3 \end{pmatrix}$$

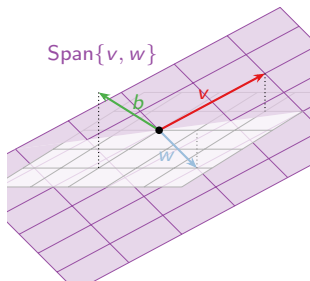
Systems of linear equations depend on the **Span** of a set of vectors!

# Spans and Solutions to Equations

## Example 2

### 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?



Columns of  $A$ :

$$v = \begin{pmatrix} 2 \\ -1 \\ 1 \end{pmatrix} \quad w = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

Output vector:

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

Is  $b$  contained in the span of the columns of  $A$ ?

# Spans and Solutions to Equations

Example 2, explained

## 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.

The first step is to put the system into an augmented matrix.

$$\left( \begin{array}{cc|c} 2 & 1 & 0 \\ -1 & 0 & 2 \\ 1 & -1 & 2 \end{array} \right) \xrightarrow{\text{row reduce}} \left( \begin{array}{cc|c} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{array} \right)$$

The last equation is  $0 = 1$ , so the system is *inconsistent*.

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.

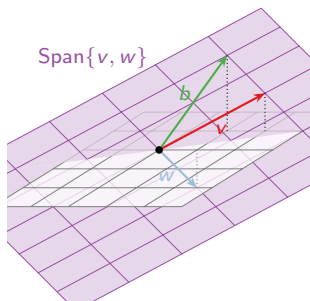


# Spans and Solutions to Equations

## Example 3

### 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?



Columns of  $A$ :

$$v = \begin{pmatrix} 2 \\ -1 \\ 1 \end{pmatrix} \quad w = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$$

Solution vector:

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

Is  $b$  contained in the span of the columns of  $A$ ?

# Spans and Solutions to Equations

Example 3, explained

## 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.

$$\left( \begin{array}{cc|c} 2 & 1 & 1 \\ -1 & 0 & -1 \\ 1 & -1 & 2 \end{array} \right) \xrightarrow{\text{row reduce}} \left( \begin{array}{cc|c} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{array} \right)$$

This gives us

$$x = 1 \quad y = -1.$$

This is consistent with the picture on the previous slide:

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

Now have four equivalent ways of writing linear systems:

1. As a *system of equations*:

$$2x_1 + 3x_2 = 7$$

$$x_1 - x_2 = 5$$

2. As an *augmented matrix*:

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

3. As a *vector equation* ( $x_1v_1 + \dots + x_nv_n = b$ ):

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

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.

## When Solutions Always Exist

**Equivalent** means that, for any given list of vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p, \mathbf{b}$ , *either all three* statements are true, *or all three* statements are false.

### Theorem

Let  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p, \mathbf{b}$  be vectors in  $\mathbf{R}^n$  and  $x_1, x_2, \dots, x_p$  be scalars. The following **are equivalent**

1. A vector  $\mathbf{b}$  is in the span of  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p$ .
2. The vector equation  $x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + \dots + x_p\mathbf{v}_p = \mathbf{b}$ , has a solution.
3. The augmented matrix below corresponds to a consistent linear system

$$\left( \mathbf{v}_1 \quad \mathbf{v}_2 \quad \dots \quad \mathbf{v}_p \mid \mathbf{b} \right).$$

### 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*.

## Why is (1) the same as (3)?

Look at **reduced echelon** forms of  $A$ .

- ▶ If  $A$  has *a pivot in each row*:

$$\begin{pmatrix} 1 & 0 & * & 0 & * \\ 0 & 1 & * & 0 & * \\ 0 & 0 & 0 & 1 & * \end{pmatrix} \quad \text{and } (A | b) \quad \begin{pmatrix} 1 & 0 & * & 0 & * & | & * \\ 0 & 1 & * & 0 & * & | & * \\ 0 & 0 & 0 & 1 & * & | & * \end{pmatrix}.$$

reduces to:

There's **no  $b$**  that makes it inconsistent, so there's *always a solution*.

- ▶ If  $A$  *doesn't have a pivot* in each row:

$$\begin{pmatrix} 1 & 0 & * & 0 & * \\ 0 & 1 & * & 0 & * \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix} \quad \text{and this can be} \quad \begin{pmatrix} 1 & 0 & * & 0 & * & | & 0 \\ 0 & 1 & * & 0 & * & | & 0 \\ 0 & 0 & 0 & 0 & 0 & | & 16 \end{pmatrix}.$$

made  
inconsistent: