## Section 1.3

**Vector Equations** 

#### Motivation

We want to think about the *algebra* in linear algebra (systems of equations and their solution sets) in terms of *geometry* (points, lines, planes, etc).

$$x - 3y = -3$$

$$2x + y = 8$$

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

To do this, we need to introduce n-dimensional space  $\mathbb{R}^n$ , and vectors inside it.

Recall that  ${\bf R}$  denotes the collection of all real numbers, i.e. the number line.

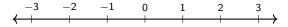
#### Definition

Let n be a positive whole number. We define

 $\mathbf{R}^n = \text{all ordered } n\text{-tuples of real numbers } (x_1, x_2, x_3, \dots, x_n).$ 

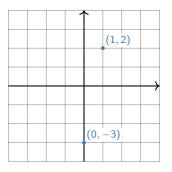
## Example

When n = 1, we just get **R** back:  $\mathbf{R}^1 = \mathbf{R}$ . Geometrically, this is the *number line*.



## Example

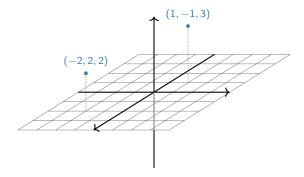
When n=2, we can think of  ${\bf R}^2$  as the *plane*. This is because every point on the plane can be represented by an ordered pair of real numbers, namely, its x- and y-coordinates.



We can use the elements of  $\mathbf{R}^2$  to *label* points on the plane, but  $\mathbf{R}^2$  is not defined to be the plane!

## Example

When n=3, we can think of  ${\bf R}^3$  as the *space* we (appear to) live in. This is because every point in space can be represented by an ordered triple of real numbers, namely, its x-, y-, and z-coordinates.



Again, we can use the elements of  $\mathbf{R}^3$  to *label* points in space, but  $\mathbf{R}^3$  is not defined to be space!

So what is  $\mathbb{R}^4$ ? or  $\mathbb{R}^5$ ? or  $\mathbb{R}^n$ ?

...go back to the *definition*: ordered *n*-tuples of real numbers

$$(x_1, x_2, x_3, \ldots, x_n).$$

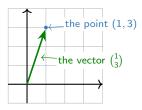
They're still "geometric" spaces, in the sense that our intuition for  $\mathbf{R}^2$  and  $\mathbf{R}^3$  sometimes extends to  $\mathbf{R}^n$ , but they're harder to visualize.

We'll make definitions and state theorems that apply to any  $\mathbf{R}^n$ , but we'll only draw pictures for  $\mathbf{R}^2$  and  $\mathbf{R}^3$ .

#### Vectors

In the previous slides, we were thinking of elements of  $\mathbb{R}^n$  as **points**: in the line, plane, space, etc.

We can also think of them as **vectors**: arrows with a given length and direction.



So the vector points *horizontally* in the amount of its x-coordinate, and *vertically* in the amount of its y-coordinate.

When we think of an element of  $\mathbf{R}^n$  as a vector, we write it as a matrix with n rows and one column:

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

We'll see why this is useful later.

#### Points and Vectors

So what is the difference between a point and a vector?

A vector need not start at the origin: *it can be located anywhere*! In other words, an arrow is determined by its length and its direction, not by its location.



These arrows all represent the vector  $\begin{pmatrix} 1 \\ 2 \end{pmatrix}$ .

However, unless otherwise specified, we'll assume a vector starts at the origin: we'll usually be sloppy and identify the vector  $\binom{1}{2}$  with the point (1,2).

This makes sense in the real world: many physical quantities, such as velocity, are represented as vectors. But it makes more sense to think of the velocity of a car as being located at the car.

Another way to think about it: a vector is a *difference* between two points, or the arrow from one point to another. (2,3)

For instance,  $\begin{pmatrix} 1 \\ 2 \end{pmatrix}$  is the arrow from (1,1) to (2,3).

## Vector Algebra

#### Definition

▶ We can add two vectors together:

$$\begin{pmatrix} a \\ b \\ c \end{pmatrix} + \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} a+x \\ b+y \\ c+z \end{pmatrix}.$$

▶ We can multiply, or **scale**, a vector by a real number c:

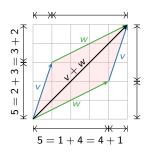
$$c\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} c \cdot x \\ c \cdot y \\ c \cdot z \end{pmatrix}.$$

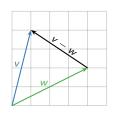
We call c a scalar to distinguish it from a vector. If v is a vector and c is a scalar, cv is called a scalar multiple of v.

(And likewise for vectors of length n.) For instance,

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} + \begin{pmatrix} 4 \\ 5 \\ 6 \end{pmatrix} = \begin{pmatrix} 5 \\ 7 \\ 9 \end{pmatrix} \quad \text{and} \quad -2 \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = \begin{pmatrix} -2 \\ -4 \\ -6 \end{pmatrix}.$$

## Vector Addition and Subtraction: Geometry





#### The parallelogram law for vector addition

Geometrically, the sum of two vectors v, w is obtained as follows: place the tail of w at the head of v. Then v+w is the vector whose tail is the tail of v and whose head is the head of w. Doing this both ways creates a parallelogram. For example,

$$\binom{1}{3} + \binom{4}{2} = \binom{5}{5}.$$

Why? The width of v + w is the sum of the widths, and likewise with the heights.

#### Vector subtraction

Geometrically, the difference of two vectors v, w is obtained as follows: place the tail of v and w at the same point. Then v - w is the vector from the head of v to the head of w. For example,

$$\begin{pmatrix} 1 \\ 4 \end{pmatrix} - \begin{pmatrix} 4 \\ 2 \end{pmatrix} = \begin{pmatrix} -3 \\ 2 \end{pmatrix}.$$

Why? If you add v - w to w, you get v.

This works in higher dimensions too!

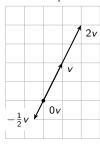


## Scalar Multiplication: Geometry

#### Scalar multiples of a vector

These have the same direction but a different length.

Some multiples of v.



$$v = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$
$$2v = \begin{pmatrix} 2 \\ 4 \end{pmatrix}$$

$$2v = \begin{pmatrix} 2 \\ 4 \end{pmatrix}$$

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

$$0v = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

All multiples of v.



So the scalar multiples of v form a *line*.

#### **Linear Combinations**

We can add and scalar multiply in the same equation:

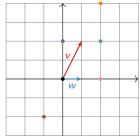
$$w = c_1 v_1 + c_2 v_2 + \cdots + c_p v_p$$

where  $c_1, c_2, \ldots, c_p$  are scalars,  $v_1, v_2, \ldots, v_p$  are vectors in  $\mathbf{R}^n$ , and w is a vector in  $\mathbf{R}^n$ .

#### Definition

We call w a linear combination of the vectors  $v_1, v_2, \ldots, v_p$ . The scalars  $c_1, c_2, \ldots, c_p$  are called the weights or coefficients.

### Example



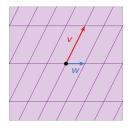
Let 
$$\mathbf{v} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$
 and  $\mathbf{w} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ .

What are some linear combinations of v and w?

- ► *v* + *w*
- V − W
- ► 2v + 0w
- ▶ 2w
- ► -v

Poll Is there any vector in  $\mathbb{R}^2$  that is *not* a linear combination of v and w?

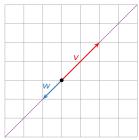
No: in fact, every vector in  $\mathbf{R}^2$  is a combination of v and w.



(The purple lines are to help measure  $how\ much$  of v and w you need to get to a given point.)

## More Examples





What are some linear combinations of  $v = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$ ?

- $ightharpoonup \frac{3}{2}V$
- $-\frac{1}{2}v$
- **•** ...

What are all linear combinations of v?

All vectors cv for c a real number. I.e., all scalar multiples of v. These form a line.

#### Question

What are all linear combinations of

$$v = \begin{pmatrix} 2 \\ 2 \end{pmatrix}$$
 and  $w = \begin{pmatrix} -1 \\ -1 \end{pmatrix}$ ?

Answer: The line which contains both vectors.

What's different about this example and the one on the poll?

## Systems of Linear Equations

Question

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

This means: can we solve the equation

$$x \begin{pmatrix} 1 \\ 2 \\ 6 \end{pmatrix} + y \begin{pmatrix} -1 \\ -2 \\ -1 \end{pmatrix} = \begin{pmatrix} 8 \\ 16 \\ 3 \end{pmatrix}$$

where x and y are the unknowns (the coefficients)? Rewrite:

$$\begin{pmatrix} x \\ 2x \\ 6x \end{pmatrix} + \begin{pmatrix} -y \\ -2y \\ -y \end{pmatrix} = \begin{pmatrix} 8 \\ 16 \\ 3 \end{pmatrix} \quad \text{or} \quad \begin{pmatrix} x - y \\ 2x - 2y \\ 6x - y \end{pmatrix} = \begin{pmatrix} 8 \\ 16 \\ 3 \end{pmatrix}.$$

This is just a system of linear equations:

$$x - y = 8$$

$$2x - 2y = 16$$

$$6x - y = 3$$

# Systems of Linear Equations

What is the relationship between the original vectors and the matrix form of the linear equation? They have the same columns!

Shortcut: You can make the augmented matrix without writing down the system of linear equations first.

## Vector Equations and Linear Equations

#### Summary

The vector equation

$$x_1v_1+x_2v_2+\cdots+x_pv_p=b,$$

where  $v_1, v_2, \ldots, v_p, b$  are vectors in  $\mathbf{R}^n$  and  $x_1, x_2, \ldots, x_p$  are scalars, has the same solution set as the linear system with augmented matrix

$$\begin{pmatrix} | & | & & & | & | \\ v_1 & v_2 & \cdots & v_p & | & b \\ | & | & & | & | & | \end{pmatrix},$$

where the  $v_i$ 's and b are the columns of the matrix.

So we now have (at least) *two* equivalent ways of thinking about linear systems of equations:

- 1. Augmented matrices.
- 2. Linear combinations of vectors (vector equations).

The last one is more geometric in nature.

It is important to know what are all linear combinations of a set of vectors  $v_1, v_2, \dots, v_p$  in  $\mathbb{R}^n$ : it's exactly the collection of all b in  $\mathbb{R}^n$  such that the vector equation (in the unknowns  $x_1, x_2, \dots, x_p$ )

$$x_1v_1+x_2v_2+\cdots+x_pv_p=b$$

has a solution (i.e., is consistent).

#### Definition

"the set of"

"such that"

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

Span
$$\{v_1, v_2, \dots, v_p\} = \{x_1v_1 + x_2v_2 + \dots + x_pv_p \mid x_1, x_2, \dots, x_p \text{ in } \mathbf{R} \}.$$

Synonyms: Span $\{v_1, v_2, \dots, v_p\}$  is the subset spanned by or generated by  $V_1, V_2, \ldots, V_D$ .

This is the first of several definitions in this class that you simply must learn. I will give you other ways to think about Span, and ways to draw pictures, but this is the definition. Having a vague idea what Span means will not help you solve any exam problems!

Now we have several equivalent ways of making the same statement:

- 1. A vector b is in the span of  $v_1, v_2, \ldots, v_p$ .
- 2. The linear system with augmented matrix

is consistent.

3. The vector equation

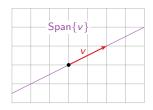
$$x_1v_1+x_2v_2+\cdots+x_pv_p=b$$

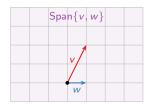
has a solution.

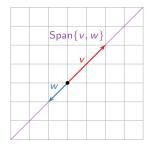
Note: **equivalent** means that, for any given list of vectors  $v_1, v_2, \ldots, v_p, b$ , *either* all three statements are true, *or* all three statements are false.

## Pictures of Span

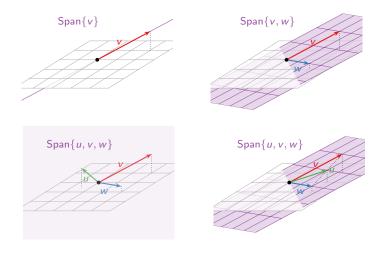
Drawing a picture of Span $\{v_1, v_2, \dots, v_p\}$  is the same as drawing a picture of all linear combinations of  $v_1, v_2, \dots, v_p$ .

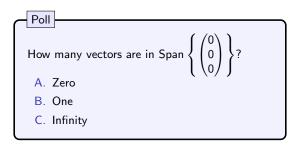






# Pictures of Span $_{\text{In }R^3}$





In general, it appears that  $Span\{v_1, v_2, \ldots, v_p\}$  is the smallest "linear space" (line, plane, etc.) containing the origin and all of the vectors  $v_1, v_2, \ldots, v_p$ .

We will make this precise later.