

# Vectors



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**Vector:** A vector is an entity with **direction** and **length** (or magnitude!)

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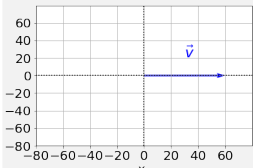
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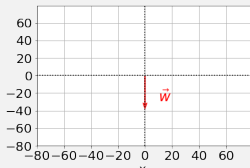
## Example

A car traveling 60 miles per hour directly East.



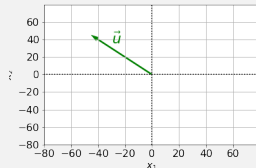
$$\mathbf{v} = \vec{v} = \begin{bmatrix} 60 \\ 0 \end{bmatrix}$$

A car traveling 40 miles per hour directly South.



$$\mathbf{w} = \vec{w} = \begin{bmatrix} 0 \\ -40 \end{bmatrix}$$

A car traveling 64 miles per hour directly North West.



$$\mathbf{u} = \vec{u} = \begin{bmatrix} -\frac{64}{\sqrt{2}} \\ \frac{64}{\sqrt{2}} \end{bmatrix}$$

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If the vector  $\mathbf{v}$  has  $n$  different real numbers, then we say that  $\mathbf{v} \in \mathbb{R}^n$ , and

$$\mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$

# Vector Equality

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- ▶ They have the **same direction and length** or
- ▶ All entries of the two vectors are the same:

$$v_k = w_k \text{ for all } 1 \leq k \leq n$$

# Vector Arithmetic (Addition)

## Definition

We define adding two vectors  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^n$  as  $\mathbf{u} + \mathbf{v} = \begin{bmatrix} u_1 + v_1 \\ \vdots \\ u_n + v_n \end{bmatrix}$ . Note: This means that we can only add vectors with the same dimensions together! And the result will have the same dimension!

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Adding the following vectors

$$\mathbf{u} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \mathbf{v} = \begin{bmatrix} 4 \\ 1 \end{bmatrix}$$

$$\mathbf{u} = \begin{bmatrix} 0 \\ -1 \\ 5 \end{bmatrix}, \mathbf{v} = \begin{bmatrix} 7 \\ -5 \\ 10 \end{bmatrix}$$

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$$\mathbf{u} + \mathbf{v} = \begin{bmatrix} 1 + 4 \\ 2 + 1 \end{bmatrix} = \begin{bmatrix} 5 \\ 3 \end{bmatrix}$$

$$\mathbf{u} + \mathbf{v} = \begin{bmatrix} 0 + 7 \\ -1 + -5 \\ 5 + 10 \end{bmatrix} = \begin{bmatrix} 7 \\ -6 \\ 15 \end{bmatrix}$$

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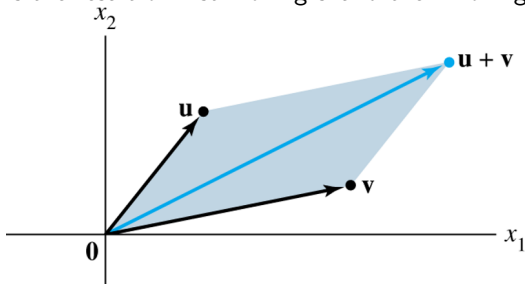
Computing the scalar product of the following

$$c = -3, \mathbf{v} = \begin{bmatrix} 2 \\ -5 \end{bmatrix}, \quad c\mathbf{v} = \begin{bmatrix} -3 \cdot 2 \\ -3 \cdot -5 \end{bmatrix} = \begin{bmatrix} -6 \\ 15 \end{bmatrix}$$

# Geometric Descriptions of Vectors in $\mathbb{R}^2$

We can think of the vector  $\mathbf{v} = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \in \mathbb{R}^2$  as a directed line segment (an arrow).

Consider two vectors  $\mathbf{u}$  and  $\mathbf{v} \in \mathbb{R}^2$ . The sum  $\mathbf{u} + \mathbf{v}$  is the result of first moving  $\mathbf{u}$  and then moving  $\mathbf{v}$ .



Consider the vector  $\mathbf{u} \in \mathbb{R}^2$  and nonzero scalar  $c$ . The scalar product  $c\mathbf{u}$  gives a vector parallel to  $\mathbf{u}$  whose magnitude has been scaled by a factor of  $c$ .

- ▶ If  $c > 0$ , then  $c\mathbf{u}$  has the same direction as  $\mathbf{u}$ .
- ▶ If  $c < 0$ , then  $c\mathbf{u}$  has the opposite direction as  $\mathbf{u}$ .
- ▶ If  $|c| < 1$ , then  $c\mathbf{u}$  is a compression of  $\mathbf{u}$ .
- ▶ If  $|c| > 1$ , then  $c\mathbf{u}$  is a stretching of  $\mathbf{u}$ .

**Parallelogram Rule for Addition:** If  $\mathbf{u}$  and  $\mathbf{v}$  are in  $\mathbb{R}^2$ , then the sum  $\mathbf{u} + \mathbf{v}$  is the diagonal of the parallelogram spanned by  $\mathbf{u}$  and  $\mathbf{v}$ .

# Properties of Vector Arithmetic

The **zero vector** is defined as the vector in  $\mathbb{R}^n$  with no magnitude, and it is denoted by  $\mathbf{0}$  or  $\vec{0}$ .

As a result of this definition, it follows that  $\mathbf{0} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ .

For all  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  in  $\mathbb{R}^n$  and all scalars  $c$  and  $d$ :

- |  |  |
|--|--|
| (i) $\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$                                | (v) $c(\mathbf{u} + \mathbf{v}) = c\mathbf{u} + c\mathbf{v}$ |
| (ii) $(\mathbf{u} + \mathbf{v}) + \mathbf{w} = \mathbf{u} + (\mathbf{v} + \mathbf{w})$ | (vi) $(c + d)\mathbf{u} = c\mathbf{u} + d\mathbf{u}$         |
| (iii) $\mathbf{u} + \mathbf{0} = \mathbf{0} + \mathbf{u} = \mathbf{u}$                 | (vii) $c(d\mathbf{u}) = (cd)\mathbf{u}$                      |
| (iv) $\mathbf{u} + (-\mathbf{u}) = \mathbf{0}$   | (viii) $1\mathbf{u} = \mathbf{u}$                            |

# Linear Combinations of Vectors

## Definition

**Linear Combination:** A linear combination of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_p \in \mathbb{R}^n$  with weights  $c_1, \dots, c_p \in \mathbb{R}$  is the vector  $\mathbf{y}$  given by

$$\mathbf{y} = c_1\mathbf{v}_1 + \cdots + c_p\mathbf{v}_p = \sum_{k=1}^p c_k\mathbf{v}_k$$

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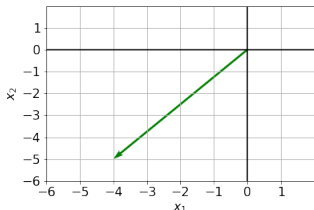
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# Giving Directions in a Grid

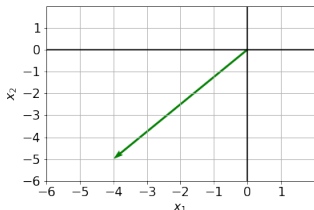


## Example

Find constants  $x_1$  and  $x_2$  such that

$$\begin{bmatrix} -4 \\ -5 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

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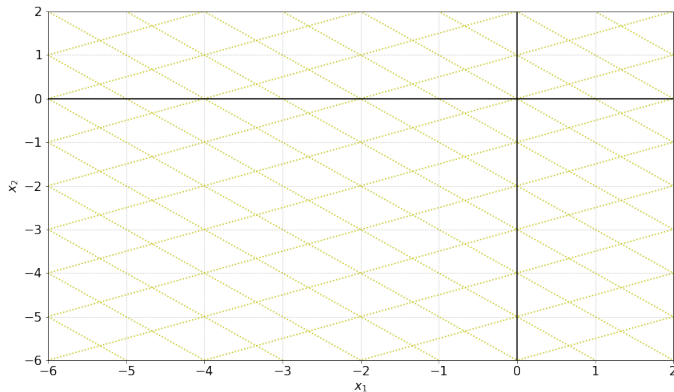
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The vectors  $e_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $e_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$  are the **standard column vectors**.

# Giving Directions in a Different System



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Find constants  $x_1$  and  $x_2$  such that

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## Solving for the Weights

First, let's simplify our statement!

$$\begin{bmatrix} -4 \\ -5 \end{bmatrix} = x_1 \begin{bmatrix} 2 \\ 1 \end{bmatrix} + x_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2x_1 - x_2 \\ x_1 + x_2 \end{bmatrix}$$

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Which becomes

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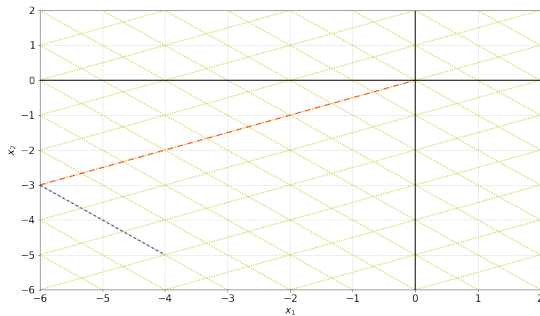
$$-5 = 1x_1 + 1x_2$$

Gaussian Elimination can solve this!

$$\left[ \begin{array}{cc|c} 2 & -1 & -4 \\ 1 & 1 & -5 \end{array} \right] \rightarrow \left[ \begin{array}{cc|c} 1 & 0 & -3 \\ 0 & 1 & -2 \end{array} \right]$$

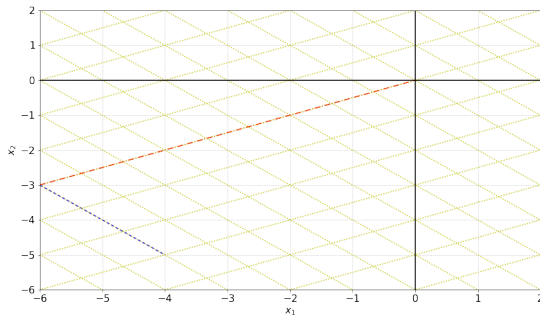
So,  $(x_1, x_2) = (-3, -2)$ . Or we can say  $\mathbf{x} = \begin{bmatrix} -3 \\ -2 \end{bmatrix}$

# What Our Directions Look Like



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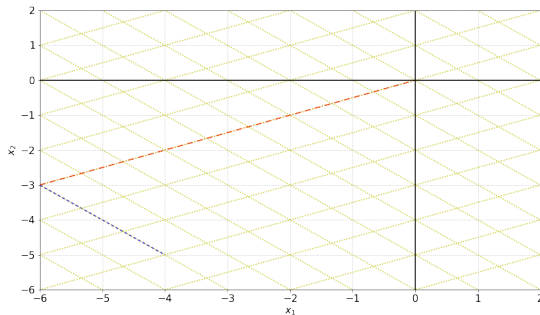


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## What Our Directions Look Like



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What about an arbitrary vector  $\mathbf{w} \in \mathbb{R}^2$ ? Can we find an  $\mathbf{x}$  such that

$$\begin{bmatrix} w_1 \\ w_2 \end{bmatrix} = x_1 \begin{bmatrix} 2 \\ 1 \end{bmatrix} + x_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

# Span of a Set of Vectors

## Definition

**Span:** Let  $\mathbf{v}_1, \dots, \mathbf{v}_p \in \mathbb{R}^n$ . Then we define  $\text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_p)$  to be the subset of  $\mathbb{R}^n$  containing all linear combinations of  $\mathbf{v}_1, \dots, \mathbf{v}_p$ .

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$$\text{Span}(\mathbf{v}_1, \dots, \mathbf{v}_p) = \left\{ \mathbf{y} \in \mathbb{R}^n \mid \mathbf{y} = \sum_{k=1}^p c_k \mathbf{v}_k \text{ for some } c_1, \dots, c_p \in \mathbb{R} \right\}$$

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## Solution

*Solve the system*

$$\left[ \begin{array}{ccc|c} \mathbf{v}_1 & \cdots & \mathbf{v}_p & \mathbf{y} \end{array} \right]$$

# Span Example

## Example

Describe the span of the vectors  $\mathbf{v}_1, \mathbf{v}_2$  given by

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \quad \mathbf{v}_2 = \begin{bmatrix} 0 \\ 2 \\ 1 \end{bmatrix}$$

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## Solution

$$c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 = c_1 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 0 \\ 2 \\ 1 \end{bmatrix} = \begin{bmatrix} c_1 \\ 2c_2 \\ c_1 + c_2 \end{bmatrix}$$

So,

$$\text{Span}(\mathbf{v}_1, \mathbf{v}_2) = \left\{ \mathbf{y} \in \mathbb{R}^3 \mid \mathbf{y} = \begin{bmatrix} c_1 \\ 2c_2 \\ c_1 + c_2 \end{bmatrix} \text{ for some } c_1, c_2 \in \mathbb{R} \right\}$$

# Span of a Set of Vectors Practice

Describe the span of the following vectors

1.  $\mathbf{v}_1 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$

2.  $\mathbf{v}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \mathbf{v}_2 = \begin{bmatrix} -4 \\ -2 \end{bmatrix}$

3.  $\mathbf{v}_1 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

4.  $\mathbf{v}_1 = \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}, \mathbf{v}_2 = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$



# Span of a Set of Vectors Practice Solutions

Describe the span of the following vectors

1.  $\mathbf{v}_1 = \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}.$

$$\text{Span}(\mathbf{v}_1) = \left\{ \mathbf{y} \in \mathbb{R}^3 \mid \mathbf{y} = \begin{bmatrix} 0 \\ c_1 \\ c_1 \end{bmatrix} \text{ for some } c_1 \in \mathbb{R} \right\}$$

2.  $\mathbf{v}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \mathbf{v}_2 = \begin{bmatrix} -4 \\ -2 \end{bmatrix}$

$$\text{Span}(\mathbf{v}_1, \mathbf{v}_2) = \left\{ \mathbf{y} \in \mathbb{R}^2 \mid \mathbf{y} = \begin{bmatrix} 2c_1 - 4c_2 \\ c_1 - 2c_2 \end{bmatrix} \text{ for some } c_1, c_2 \in \mathbb{R} \right\}$$

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$$\text{Span}(\mathbf{v}_1, \mathbf{v}_2) = \left\{ \mathbf{y} \in \mathbb{R}^3 \mid \mathbf{y} = \begin{bmatrix} c_1 \\ c_2 \\ -2c_1 - c_2 \end{bmatrix} \text{ for some } c_1, c_2 \in \mathbb{R} \right\}$$

## More Span Practice

Determine if  $y = \begin{bmatrix} 4 \\ 5 \\ -1 \end{bmatrix}$  is in  $\text{Span} \left( \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} \right)$

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Perform Gaussian Elimination on the augmented matrix

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## More Span Practice

Determine if  $y = \begin{bmatrix} 4 \\ 5 \\ -1 \end{bmatrix}$  is in  $\text{Span} \left( \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} \right)$

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No!

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Recall for the system,

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But writing this can be cumbersome!

## Using Vectors!

We recall from earlier that we can write out our answer in terms of vectors!

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## Span of a Set of Vectors Conceptual Practice

Let  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{w}$  be vectors in  $\mathbb{R}^n$ . Prove that if  $\mathbf{w} \in \text{Span}(\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3)$ , then  $\text{Span}(\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{w}) = \text{Span}(\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3)$