

# Matrix-Vector Equations

**Geometric Algorithms**  
**Lecture 5**

# Practice Problem

Is the vector  $\begin{bmatrix} 9 \\ 3 \\ -14 \end{bmatrix}$  in  $\text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 3 \\ 2 \\ -4 \end{bmatrix} \right\}$ ?

# Answer

$$\begin{bmatrix} 9 \\ 3 \\ -14 \end{bmatrix} \in \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 3 \\ 2 \\ -4 \end{bmatrix} \right\} ?$$

$$x_1 \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix} + x_2 \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix} + x_3 \begin{bmatrix} 3 \\ 2 \\ -4 \end{bmatrix} = \begin{bmatrix} 9 \\ 3 \\ -14 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & 3 & 9 \\ 0 & 1 & 2 & 3 \\ -2 & -1 & -4 & -14 \end{bmatrix}$$

$$R_3 \leftarrow R_3 + 2R_1$$

$$\rightarrow$$

$$\begin{bmatrix} 1 & 1 & 3 & 9 \\ 0 & 1 & 2 & 3 \\ 0 & 1 & 2 & 4 \end{bmatrix}$$

NO

$$R_3 \leftarrow R_3 - R_2$$

$$\rightarrow$$

$$\begin{bmatrix} 1 & 1 & 3 & 9 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

# Answer

Solve the system of linear equations with the augmented matrix

$$\left[ \begin{array}{cccc} 1 & 1 & 3 & 9 \\ 0 & 1 & 2 & 3 \\ -2 & -1 & -4 & -14 \end{array} \right]$$

# Answer

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# Answer

$$\begin{bmatrix} 1 & 1 & 3 & 9 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

no solution  $\equiv$  not in the span

# Outline

- » Motivate the study of matrix–vector equations
- » Formally define matrix–vector multiplication
- » Revisit spans
- » Take stock of our perspectives on systems of linear equations



# Keywords

matrix-vector multiplication

the matrix equation

inner-product

row-column rule

# Recap

**Recall: Vector "Interface"**

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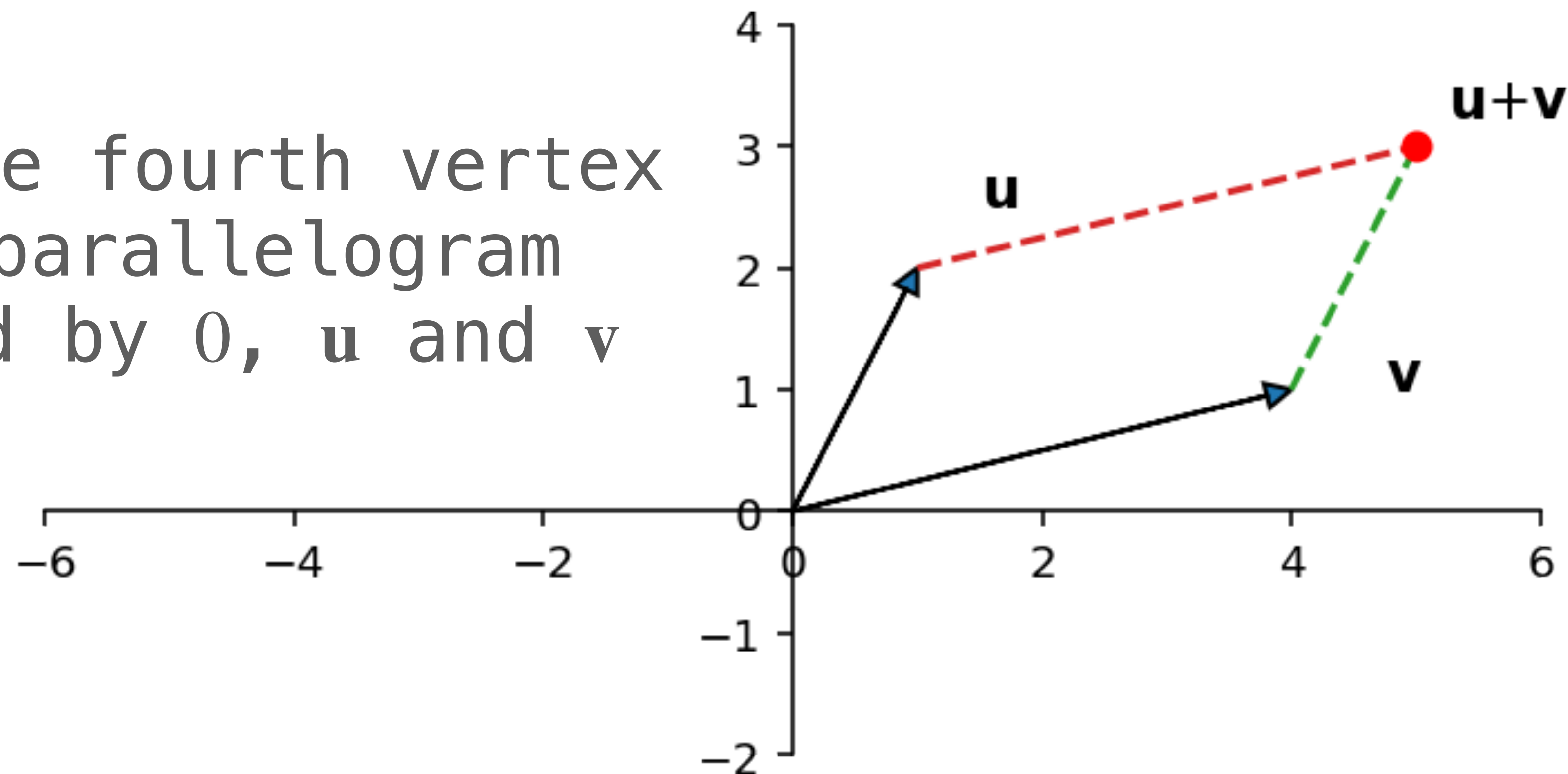
**scaling**      what does  $a\mathbf{v}$  (multiplying a vector by a real number) mean?

What properties do they need to satisfy?

# Recall: Vector Addition (Geometrically)

in  $\mathbb{R}^2$  it's called the *parallelogram rule*

$\mathbf{u} + \mathbf{v}$  is the fourth vertex  
of the parallelogram  
generated by  $\mathbf{0}$ ,  $\mathbf{u}$  and  $\mathbf{v}$

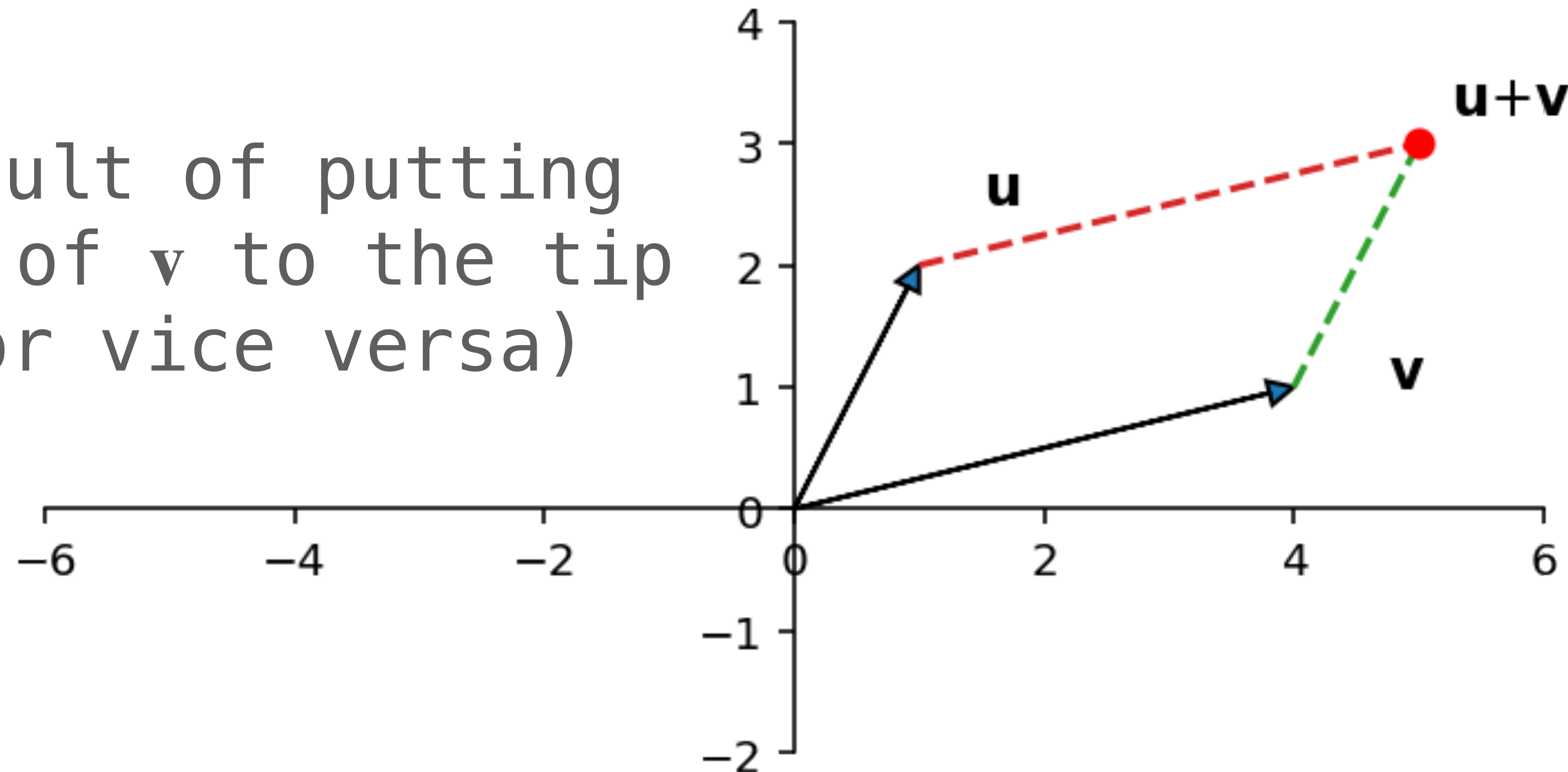




# Vector Addition (Geometrically)

or the *tip-to-tail rule*

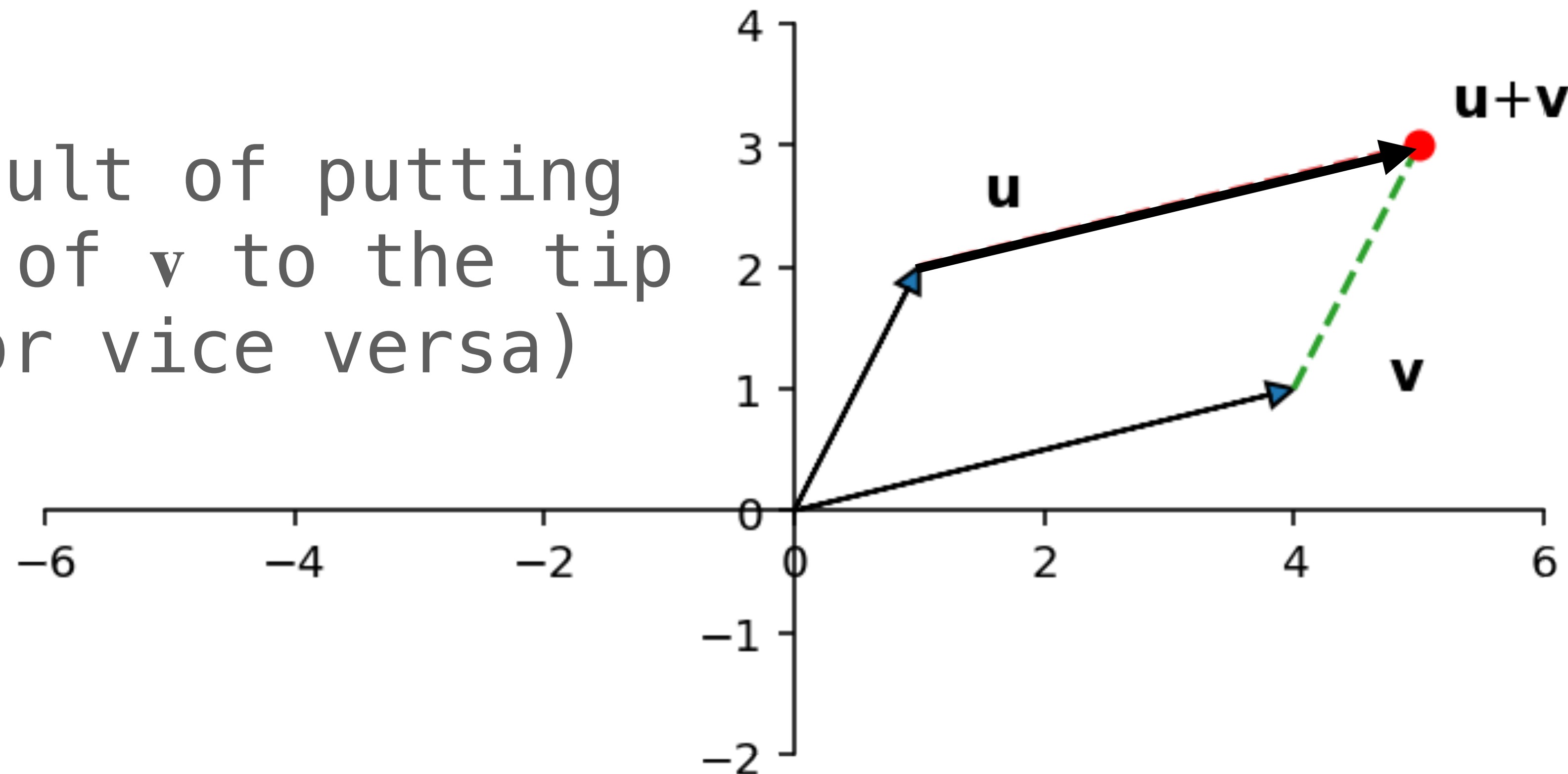
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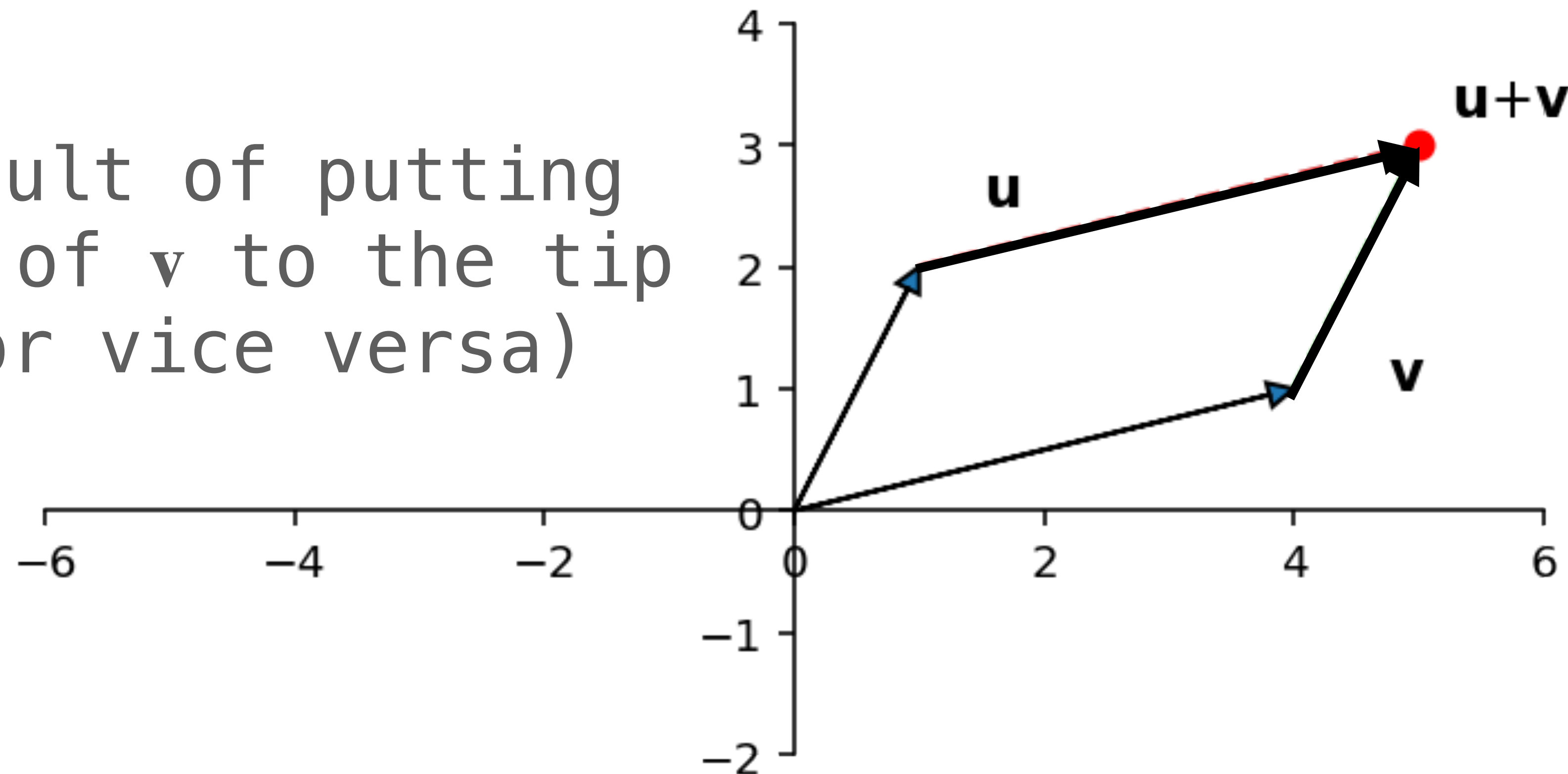
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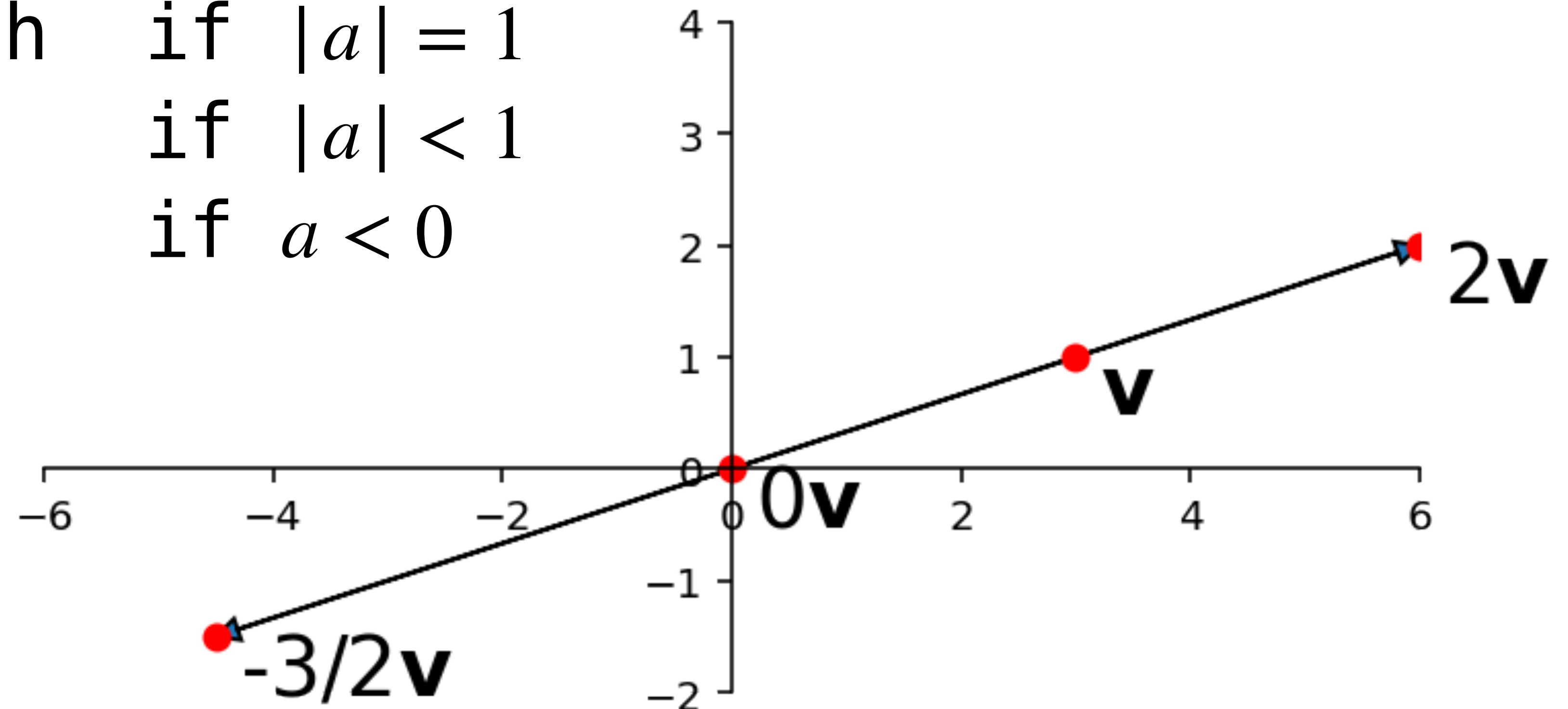
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# Recall Vector Scaling (Geometrically)

longer	if $ a  > 1$
the same length	if $ a  = 1$
shorter	if $ a  < 1$
reversed	if $a < 0$



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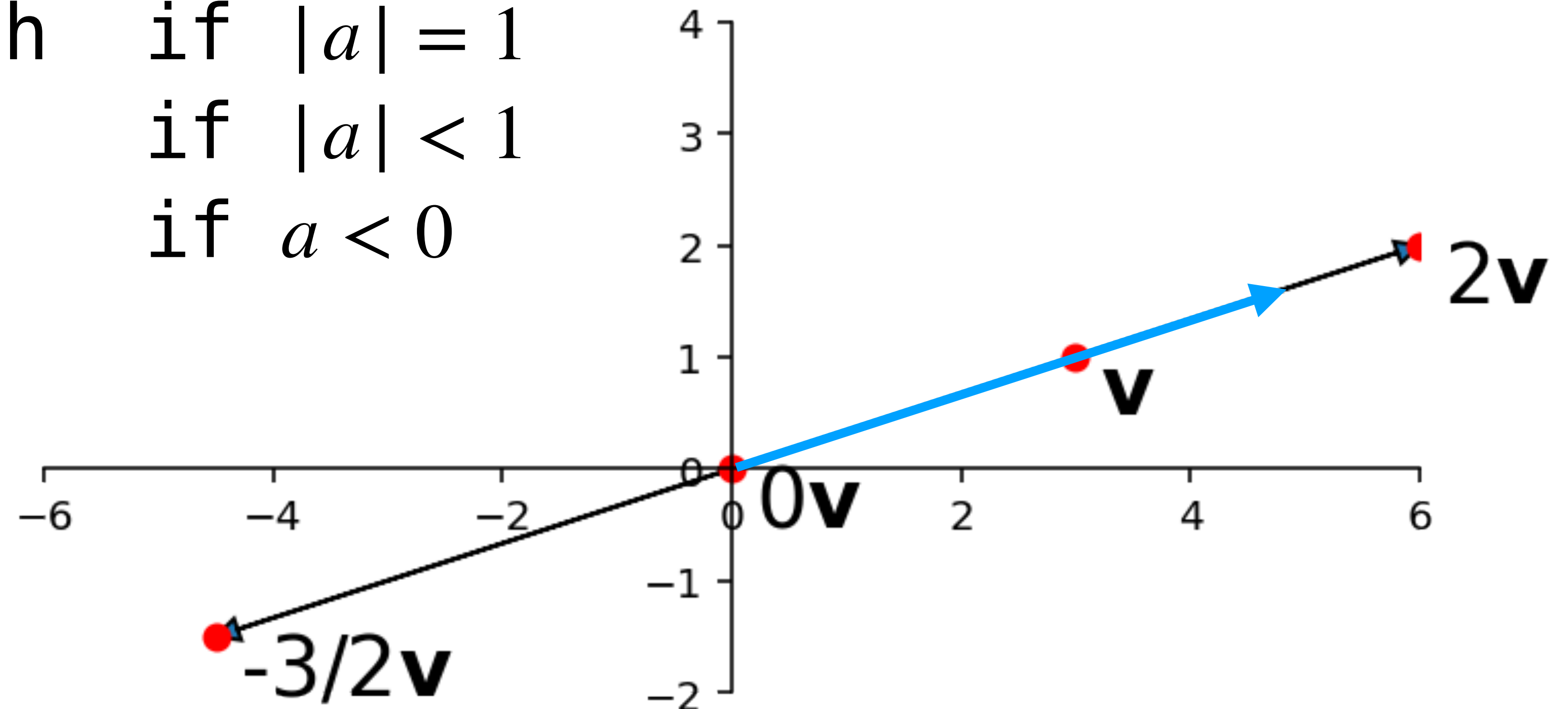
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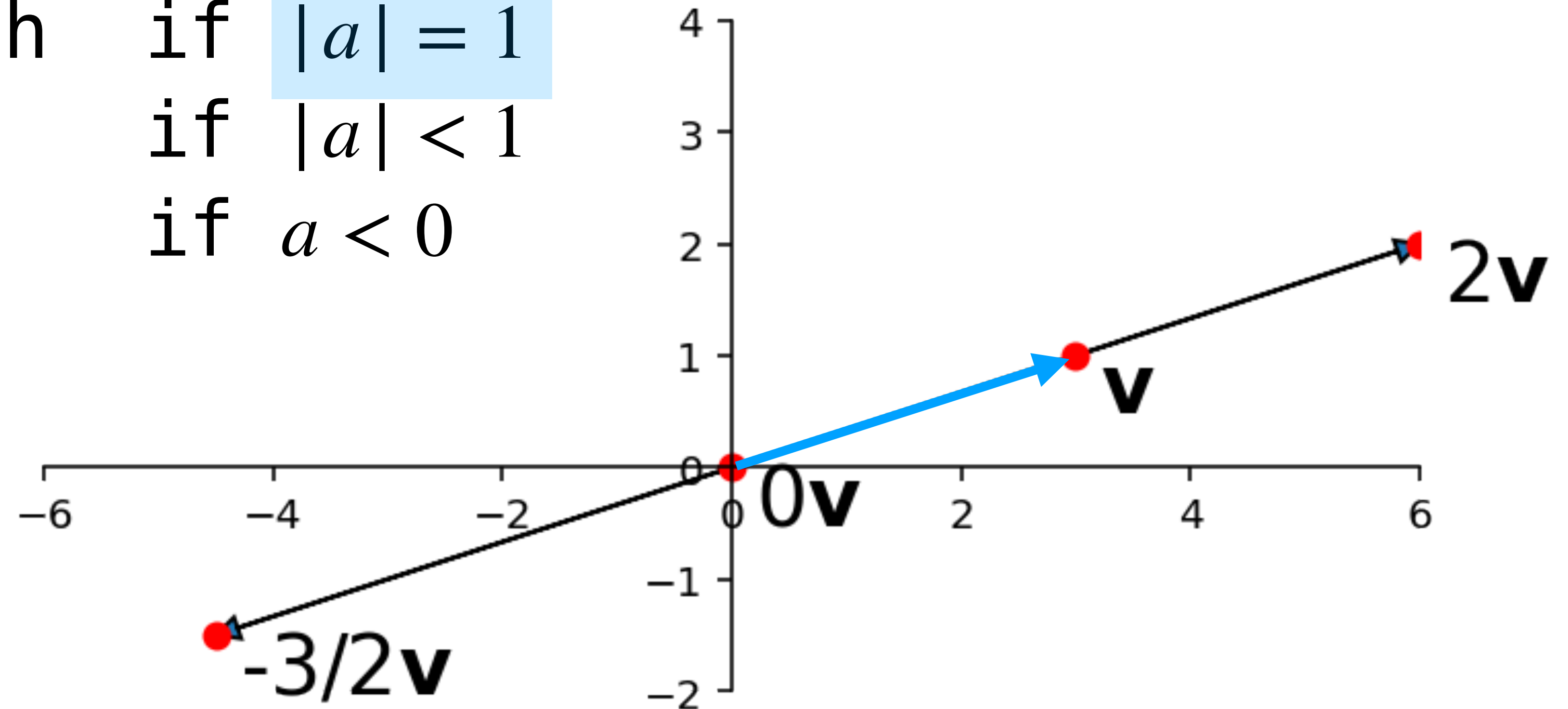
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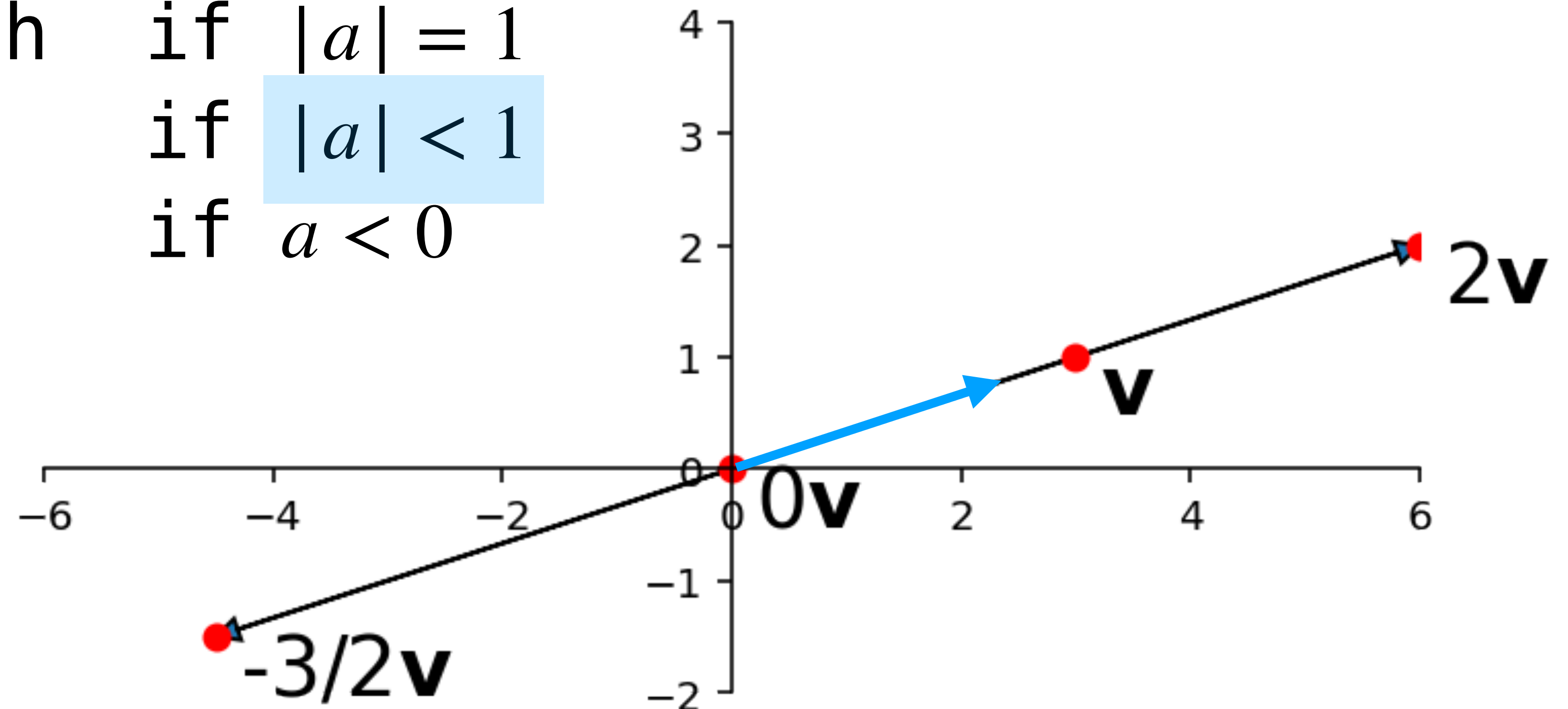
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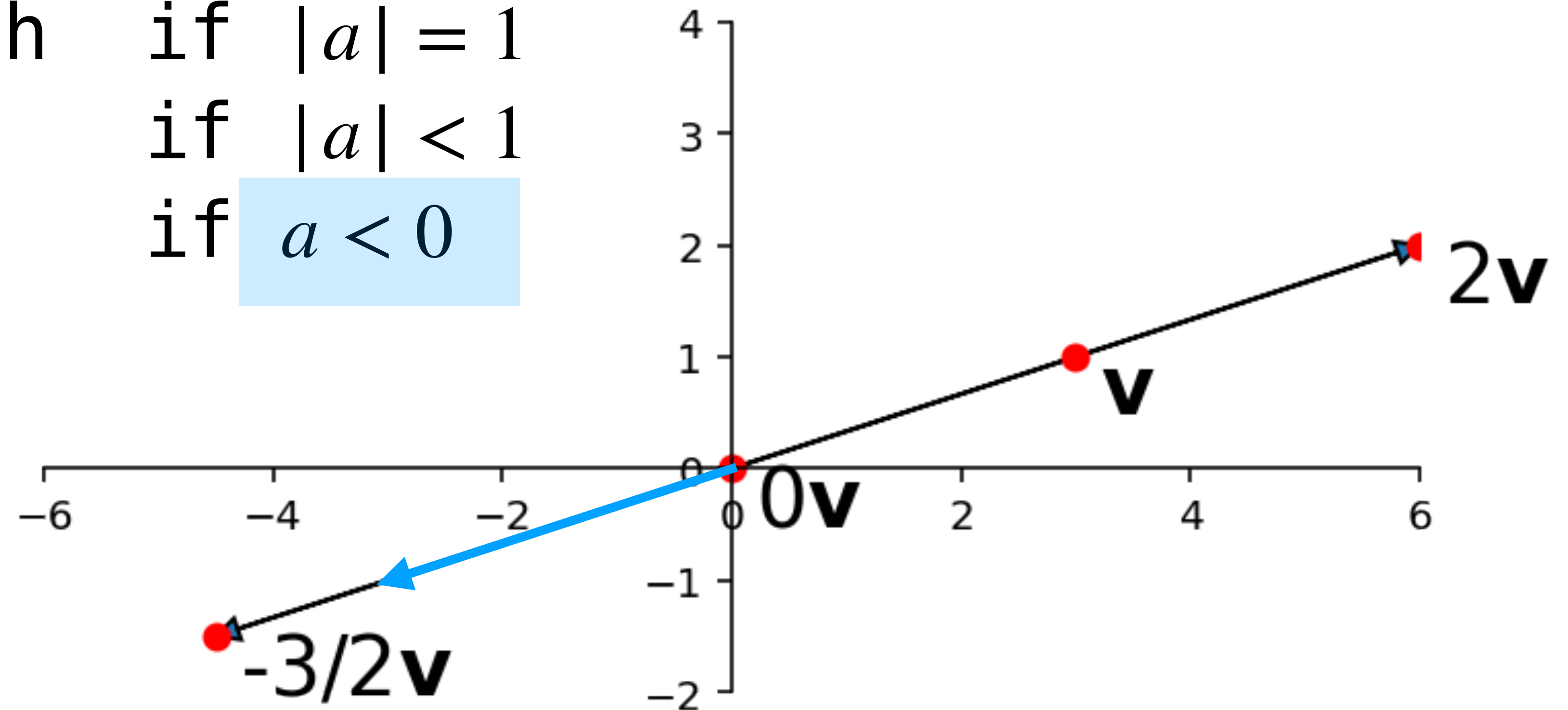
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# Recall: Linear Combinations

**Definition.** a *linear combination* of vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  is a vector of the form

$$\alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots + \alpha_n \mathbf{v}_n$$

where  $\alpha_1, \alpha_2, \dots, \alpha_n$  are in  $\mathbb{R}$

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where  $\alpha_1, \alpha_2, \dots, \alpha_n$  are in  $\mathbb{R}$   
weights

# Recall: Linear Combinations (Example)

$$3 \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} + 2 \begin{bmatrix} 2 \\ 0 \\ 3 \\ -1 \end{bmatrix} - \begin{bmatrix} -3 \\ 4 \\ 2 \\ 0 \end{bmatrix}$$

# Recall: The Fundamental Concern

Can  $\mathbf{u}$  be written as a linear combination of

$\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ ?

That is, are there weights  $\alpha_1, \alpha_2, \dots, \alpha_n$  such that

$$\alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots \alpha_n \mathbf{v}_n = \mathbf{u}?$$

# Recall: The Fundamental Connection

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} & b_1 \\ a_{21} & a_{22} & \cdots & a_{2n} & b_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} & b_m \end{bmatrix}$$

augmented matrix

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system of linear equations

$$x_1 \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix} + x_2 \begin{bmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix} + \cdots + x_n \begin{bmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

vector equation

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this is our big  
shift in  
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# Motivation

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augmented matrix

Why not view  
these as a  
vector too?

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vector equation

# Solutions as Vectors

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so it can be represented as a vector

Can we view a linear system as a single equation  
with matrices and vectors?

How do matrices and vectors "interface"?

# Matrix-Vector Multiplication

# Matrix-Vector "Interface"

multiplication

what does  $A\mathbf{v}$  mean when  $A$  is a matrix and  $\mathbf{v}$  is a vector?

# Matrix-Vector Multiplication (Pictorially)

$AS$

# Matrix-Vector Multiplication (Pictorially)

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix}$$

# Matrix-Vector Multiplication (Pictorially)

The diagram illustrates matrix-vector multiplication pictorially. On the left, a matrix is represented by a large left square bracket followed by four vertical columns. The first column contains  $a_{11}$ ,  $a_{21}$ , a vertical ellipsis, and  $a_{m1}$ . The second column contains  $a_{12}$ ,  $a_{22}$ , a vertical ellipsis, and  $a_{m2}$ . This is followed by three dots, then a third column containing  $a_{1n}$ ,  $a_{2n}$ , a vertical ellipsis, and  $a_{mn}$ . The matrix part is closed by a large right square bracket. To the right of the matrix is a vector, represented by a large left square bracket followed by four light blue rectangular boxes. The first box contains  $s_1$ , the second  $s_2$ , the third a vertical ellipsis, and the fourth  $s_n$ . The vector part is closed by a large right square bracket.

$$\left[ \begin{array}{c} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{array} \begin{array}{c} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{array} \dots \begin{array}{c} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{array} \right] \left[ \begin{array}{c} s_1 \\ s_2 \\ \vdots \\ s_n \end{array} \right]$$

# Matrix-Vector Multiplication (Pictorially)

$$s_1 \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix} + s_2 \begin{bmatrix} a_{21} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix} + \dots + s_n \begin{bmatrix} a_{m1} \\ a_{m2} \\ \vdots \\ a_{mn} \end{bmatrix}$$

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a linear combination of the columns where  
s defines the weights



# Why keeping track of matrix size is important

this only works if the number of *columns* of the matrix matches the number of *rows* of the vector

$$\begin{bmatrix} * & \dots & * \\ * & \dots & * \\ \vdots & \ddots & \vdots \\ * & \dots & * \\ * & \dots & * \end{bmatrix} \begin{bmatrix} * \\ \vdots \\ * \end{bmatrix} = \begin{bmatrix} * \\ * \\ \vdots \\ * \\ * \end{bmatrix}$$

$(m \times n)$        $(n \times 1)$        $(m \times 1)$

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$$\begin{array}{ccc} \begin{array}{c} \textcolor{blue}{m} \end{array} \begin{array}{c} \textcolor{red}{n} \\ \left[ \begin{array}{ccc} * & \dots & * \\ * & \dots & * \\ \vdots & \ddots & \vdots \\ * & \dots & * \\ * & \dots & * \end{array} \right] \end{array} & \begin{array}{c} \textcolor{red}{n} \begin{array}{c} \textcolor{purple}{1} \\ \left[ \begin{array}{c} * \\ \vdots \\ * \end{array} \right] \end{array} & = & \begin{array}{c} \textcolor{blue}{m} \begin{array}{c} \textcolor{purple}{1} \\ \left[ \begin{array}{c} * \\ * \\ \vdots \\ * \\ * \end{array} \right] \end{array} \end{array} \\ \begin{array}{c} (\textcolor{blue}{m} \times \textcolor{red}{n}) \end{array} & \begin{array}{c} (\textcolor{red}{n} \times \textcolor{purple}{1}) \end{array} & & \begin{array}{c} (\textcolor{blue}{m} \times \textcolor{purple}{1}) \end{array} \end{array}$$

# Non-Example

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 3 \end{bmatrix} + 2 \begin{bmatrix} 2 \\ 4 \end{bmatrix} + 3 ???$$

# Non-Example

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# Non-Example

The diagram illustrates a matrix multiplication and a subsequent scalar multiplication. On the left, a  $(2 \times 2)$  matrix (represented by two red vertical rectangles) is multiplied by a  $(3 \times 1)$  vector (represented by a blue vertical rectangle). The dimensions are indicated below the matrices. The result is shown as an equals sign followed by three terms: a scalar (blue square) multiplied by a  $(2 \times 1)$  vector (red rectangle), plus another scalar (blue square) multiplied by a  $(2 \times 1)$  vector (red rectangle), plus a third scalar (blue square) followed by three question marks. This represents a non-example of scalar multiplication where the scalar is a single element from the first vector.

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = 1 \begin{bmatrix} 1 \\ 3 \end{bmatrix} + 2 \begin{bmatrix} 2 \\ 4 \end{bmatrix} + 3 ???$$

$(2 \times 2) \quad (3 \times 1)$

# Non-Example

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THESE DON'T MATCH

$(2 \times 2)$   $(3 \times 1)$

# Example

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THESE MATCH

$(2 \times 2)$   $(2 \times 1)$

# Matrix-Vector Multiplication

**Definition.** Given a  $(m \times n)$  matrix  $A$  with columns  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ , and a vector  $\mathbf{v}$  in  $\mathbb{R}^n$ , we define

$$A\mathbf{v} = [\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n] \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} = \underbrace{v_1\mathbf{a}_1 + v_2\mathbf{a}_2 + \dots + v_n\mathbf{a}_n}_{\text{lin. comb.}}$$

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$\mathbb{R}^{n \times 1}$  = set of  $n \times 1$  matrices  
 $\mathbb{R}^{n \times 1}$  = set of  $n$ -element vectors

Euclidean Space

# Matrix-Vector Multiplication

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$A\mathbf{v}$  is a linear combination of the columns of  $A$  with weights given by  $\mathbf{v}$

# Algebraic Properties

The algebraic properties of matrix–vector multiplication are **very important**.

$$1. \quad A(\mathbf{u} + \mathbf{v}) = A\mathbf{u} + A\mathbf{v}$$

$$2. \quad A(c\mathbf{v}) = c(A\mathbf{v})$$

# Algebraic Properties

The algebraic properties of matrix–vector multiplication are **very important**.

$$1. \quad A(\mathbf{u} + \mathbf{v}) = A\mathbf{u} + A\mathbf{v}$$

(additivity)

$$2. \quad A(c\mathbf{v}) = c(A\mathbf{v})$$

(homogeneity)

There are only two, please memorize them...

# Derivation of (1) for $A$ in $\mathbb{R}^{n \times 3}$

$$A \left( \vec{u} + \vec{v} \right)$$
$$[\mathbf{a}_1 \quad \mathbf{a}_2 \quad \mathbf{a}_3] \left( \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} + \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} \right)$$



# Derivation of (1) for $A$ in $\mathbb{R}^{n \times 3}$

$$[\mathbf{a}_1 \quad \mathbf{a}_2 \quad \mathbf{a}_3] \begin{pmatrix} \begin{bmatrix} u_1 + v_1 \\ u_2 + v_2 \\ u_3 + v_3 \end{bmatrix} \end{pmatrix}$$

by vector addition

# Derivation of (1) for $A$ in $\mathbb{R}^{n \times 3}$

$$(u_1 + v_1)\mathbf{a}_1 + (u_2 + v_2)\mathbf{a}_2 + (u_3 + v_3)\mathbf{a}_3$$

by matrix vector multiplication

**Derivation of (1) for  $A$  in  $\mathbb{R}^{n \times 3}$**

$$(5+2) \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = 5 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} + 2 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$u_1 \mathbf{a}_1 + v_1 \mathbf{a}_1 + u_2 \mathbf{a}_2 + v_2 \mathbf{a}_2 + u_3 \mathbf{a}_3 + v_3 \mathbf{a}_3$$

by vector scaling (distribution)

# Derivation of (1) for $A$ in $\mathbb{R}^{n \times 3}$

$$(u_1 \mathbf{a}_1 + u_2 \mathbf{a}_2 + u_3 \mathbf{a}_3) + (v_1 \mathbf{a}_1 + v_2 \mathbf{a}_2 + v_3 \mathbf{a}_3)$$

by rearranging

# Derivation of (1) for $A$ in $\mathbb{R}^{n \times 3}$

$$\begin{matrix} A & \overset{\sim}{u} & + & A & \overset{\sim}{v} \\ [a_1 & a_2 & a_3] & \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} & + & [a_1 & a_2 & a_3] & \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} \end{matrix}$$

by matrix vector multiplication

# Derivation of (1) for $A$ in $\mathbb{R}^{n \times 3}$

$$[\mathbf{a}_1 \quad \mathbf{a}_2 \quad \mathbf{a}_3] \left( \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} + \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix} \right)$$

equals

$$[\mathbf{a}_1 \quad \mathbf{a}_2 \quad \mathbf{a}_3] \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} + [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \mathbf{a}_3] \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}$$

fin

# A Common Error

$$Av \neq vA$$

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It's **important** that we write our matrix-vectors multiplications with the matrix on the left



# A Common Error

$$Av \neq vA$$

It's **important** that we write our matrix–vectors multiplications with the matrix on the left

This may feel artificial now, since the RHS is meaningless to us now, but it won't be for long

# Looking forward a bit

$$\begin{bmatrix} * & \dots & * \\ * & \dots & * \\ \vdots & \ddots & \vdots \\ * & \dots & * \\ * & \dots & * \end{bmatrix} \begin{bmatrix} * \\ \vdots \\ * \end{bmatrix} = \begin{bmatrix} * \\ * \\ \vdots \\ * \\ * \end{bmatrix}$$

**Remember.** column vectors are matrices with 1 column

Eventually we'll be able to view all of these as  
*matrix operations*

# Question

$$\begin{bmatrix} 2 & -3 & 4 \\ -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 5 \\ 5 \\ 4 \end{bmatrix}$$

*Compute the above matrix-vector multiplication*

$$5 \begin{bmatrix} 2 \\ -1 \end{bmatrix} + 5 \begin{bmatrix} -3 \\ 1 \end{bmatrix} + 4 \begin{bmatrix} 4 \\ 0 \end{bmatrix} = \begin{bmatrix} 10 \\ -5 \end{bmatrix} + \begin{bmatrix} -15 \\ 5 \end{bmatrix} + \begin{bmatrix} 16 \\ 0 \end{bmatrix} = \begin{bmatrix} 11 \\ 0 \end{bmatrix}$$

**Answer**

$$\begin{bmatrix} 2 & -3 & 4 \\ -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 5 \\ 5 \\ 4 \end{bmatrix}$$

# Answer

$$\begin{bmatrix} 2 & -3 & 4 \\ -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 5 \\ 5 \\ 4 \end{bmatrix}$$

# Answer

$$5 \begin{bmatrix} 2 \\ -1 \end{bmatrix} + 5 \begin{bmatrix} -3 \\ 1 \end{bmatrix} + 4 \begin{bmatrix} 4 \\ 0 \end{bmatrix}$$

# Answer

$$\begin{bmatrix} 10 \\ -5 \end{bmatrix} + \begin{bmatrix} -15 \\ 5 \end{bmatrix} + \begin{bmatrix} 16 \\ 0 \end{bmatrix}$$

# Answer

$$\begin{bmatrix} 1 & 1 \\ 0 \end{bmatrix}$$



# A Tip

$$\begin{bmatrix} 2 & -3 & 4 \\ -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 5 \\ 5 \\ 4 \end{bmatrix} = \begin{bmatrix} ? \\ ? \end{bmatrix}$$

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$$\begin{bmatrix} 2 & -3 & 4 \\ -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 5 \\ 5 \\ 4 \end{bmatrix} = \begin{bmatrix} ? \\ ? \end{bmatrix}$$

$$5(2) + 5(-3) + 4(4) = 11$$

# A Tip

$$\begin{bmatrix} 2 & -3 & 4 \\ -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 5 \\ 5 \\ 4 \end{bmatrix} = \begin{bmatrix} 11 \\ ? \end{bmatrix}$$

$$5(-1) + 5(1) + 4(0) = 0$$

# A Tip

$$\begin{bmatrix} 2 & -3 & 4 \\ -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 5 \\ 5 \\ 4 \end{bmatrix} = \begin{bmatrix} 11 \\ 0 \end{bmatrix}$$

# Calculating $A\mathbf{v}$

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix} = \begin{bmatrix} ? \\ ? \\ \vdots \\ ? \end{bmatrix}$$

$$v_1 = a_{11}s_1 + a_{12}s_2 + \dots + a_{1n}s_n = \sum_{i=1}^n a_{1i}s_i$$

# Calculating $A\mathbf{v}$

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix} = \begin{bmatrix} v_1 \\ ? \\ \vdots \\ ? \end{bmatrix}$$

$$v_2 = a_{21}s_1 + a_{22}s_2 + \dots + a_{2n}s_n = \sum_{i=1}^n a_{2i}s_i$$

# Calculating $A\mathbf{v}$

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ ? \end{bmatrix}$$

$$v_m = a_{m1}s_1 + a_{m2}s_2 + \cdots + a_{mn}s_n = \sum_{i=1}^n a_{mi}s_i$$

# Calculating $A\mathbf{v}$

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_m \end{bmatrix}$$



# Row-Column Rule

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^n a_{1i}s_i \\ \sum_{i=1}^n a_{2i}s_i \\ \vdots \\ \sum_{i=1}^n a_{mi}s_i \end{bmatrix}$$

Inner product:

$$\underbrace{[a_1 \ a_2 \ \cdots \ a_n]}_{1 \times n} \underbrace{\begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix}}_{n \times 1} = \sum_{i=1}^n a_i s_i \in \mathbb{R}$$

$1 \times n$        $n \times 1$        $1 \times 1$

# Inner Product

$$\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

**Definition.** The **inner product** of vectors  $\mathbf{u}$  and  $\mathbf{v}$  in  $\mathbb{R}^n$  is defined the

$$\langle \mathbf{u}, \mathbf{v} \rangle = \mathbf{u} \cdot \mathbf{v} = \sum_{i=1}^n u_i v_i$$

$$\left\langle \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix} \right\rangle = 1(4) + 2(5) + 3(6) = 32$$

# Row-Column Rule

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^n a_{1i}s_i \\ \sum_{i=1}^n a_{2i}s_i \\ \vdots \\ \sum_{i=1}^n a_{mi}s_i \end{bmatrix}$$

Inner product:  $[a_1 \ a_2 \ \cdots \ a_n] \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix} = \sum_{i=1}^n a_i s_i$

# Row-Column Rule

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The  $i$ th entry of the  $As$  is the inner product of the  $i$ th row of  $A$  and  $s$

# The Matrix Equation

# Recall: Vector Equations

$$x_1 \mathbf{a}_1 + x_2 \mathbf{a}_2 + \dots + x_n \mathbf{a}_n = \mathbf{b}$$

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**Question.** Can  $\mathbf{b}$  be written as a linear combination of  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ ?

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**The Idea.** think of the weights as *unknowns*



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**The Idea.** think of the weights as *unknowns*

we can use the same idea for matrix–vector  
multiplication

# The Matrix Equation

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Can  $\mathbf{b}$  be written as a linear combination of **the columns of  $A$** ?

# The Matrix Equation

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Can  $\mathbf{b}$  be written as a linear combination of **the columns of  $A$** ?

**The Idea.** write the "vector part" of our matrix-vector multiplication as an *unknown*

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

# How To: The Matrix Equation

**Question.** Does  $A\mathbf{x} = \mathbf{b}$  have a solution?

**Question.** Is  $A\mathbf{x} = \mathbf{b}$  consistent?

**Question.** Write down a solution to the equation  $A\mathbf{x} = \mathbf{b}$

# How To: The Matrix Equation

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(augmented matrix)

$$[\mathbf{a}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_n \quad \mathbf{b}]$$

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(augmented matrix)

$$[\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n \ \mathbf{b}]$$

**!!they all have the same solution set!!**

# HOW TO: The Matrix Equation

**Question.** Write down a solution to the equation  $A\mathbf{x} = \mathbf{b}$

**Solution.**

Use Gaussian elimination (or other means) to convert  $[\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n \ \mathbf{b}]$  to reduced echelon form

Then read off a solution from the reduced echelon form

**Full Span**

**Recall: Span**

# Recall: Span

**Definition.** the *span* of a set of vectors is the set of all possible linear combinations of them

$$\text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\} = \{\alpha_1\mathbf{v}_1 + \alpha_2\mathbf{v}_2 + \dots\alpha_n\mathbf{v}_n : \alpha_1, \alpha_2, \dots, \alpha_n \text{ are in } \mathbb{R}\}$$



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$\mathbf{u} \in \text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  exactly when  $\mathbf{u}$  can be expressed as a linear combination of those vectors

# Spans (with Matrices)

**Definition.** the *span* of the vectors  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$  is:

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the span of the columns of a matrix  $A$   
is the set of vectors resulting  
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the span of the columns of a matrix  $A$   
is the set of vectors resulting  
from multiplying  $A$  by any vector

(we will soon start thinking of  $A$  as a way of *transforming* vectors)

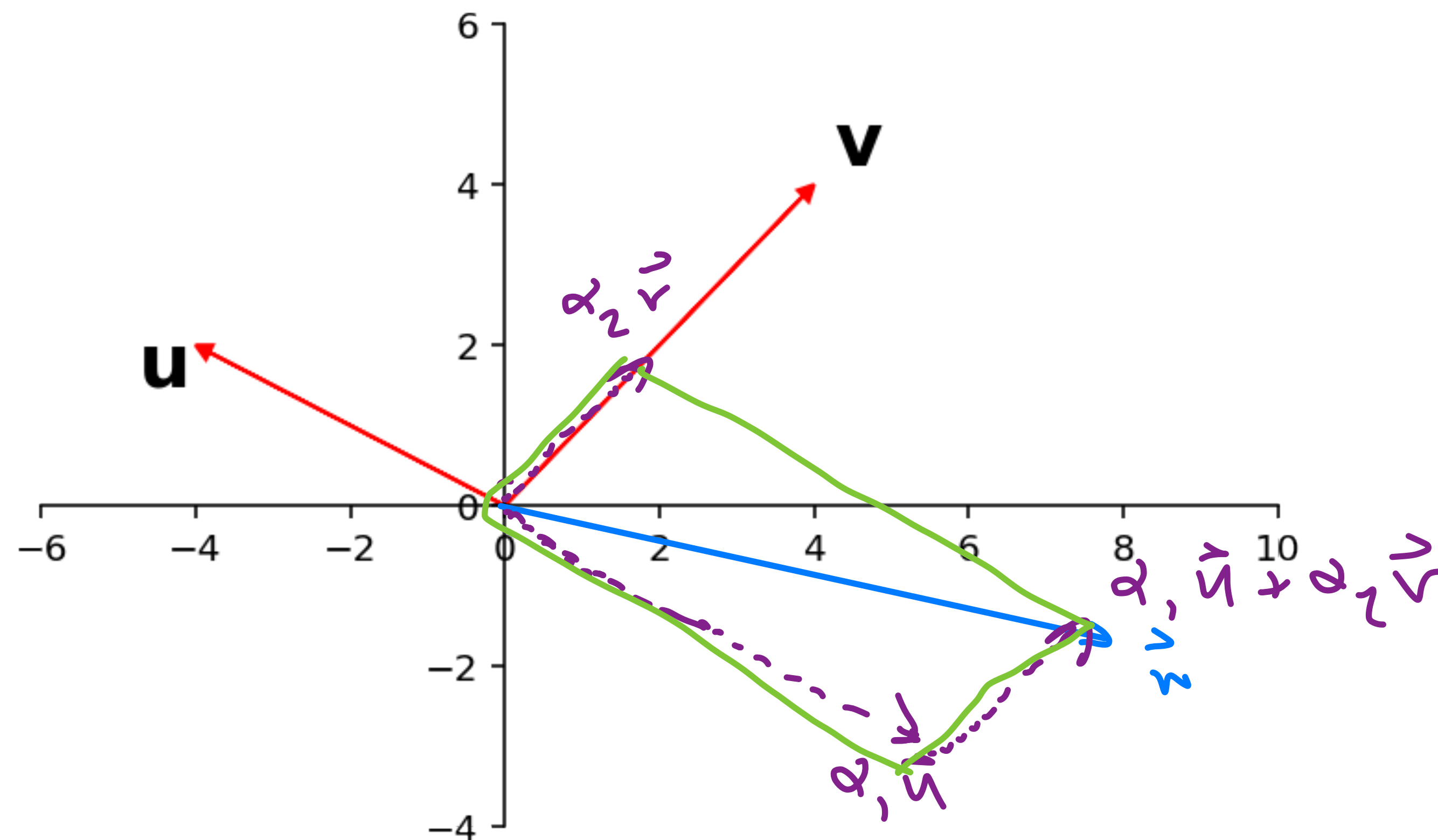
# Spanning all of $\mathbb{R}^2$

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if two (or more) vectors in  $\mathbb{R}^2$  span a plane,  
they must span all of  $\mathbb{R}^2$ . They "fill up"  $\mathbb{R}^2$

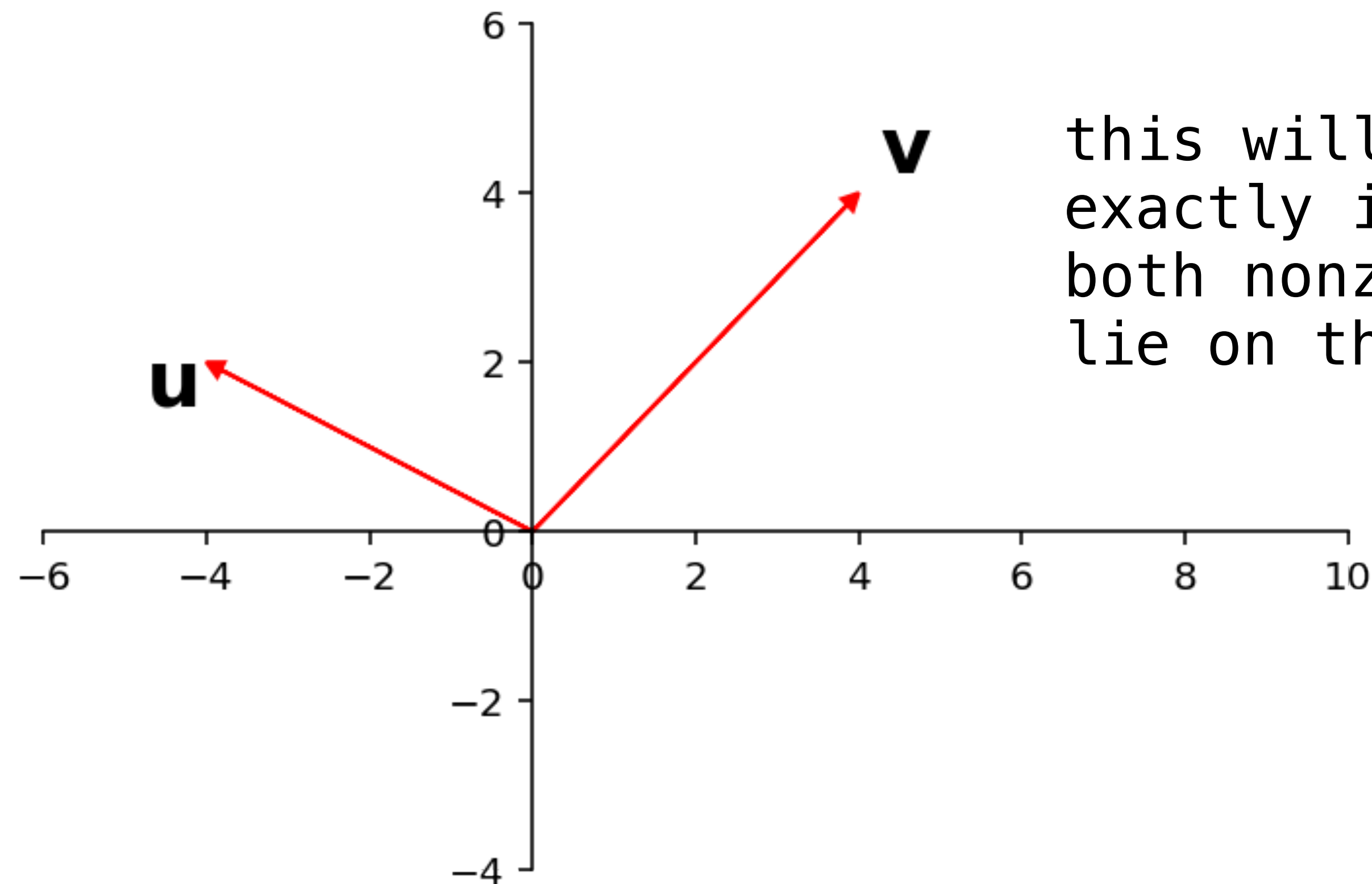
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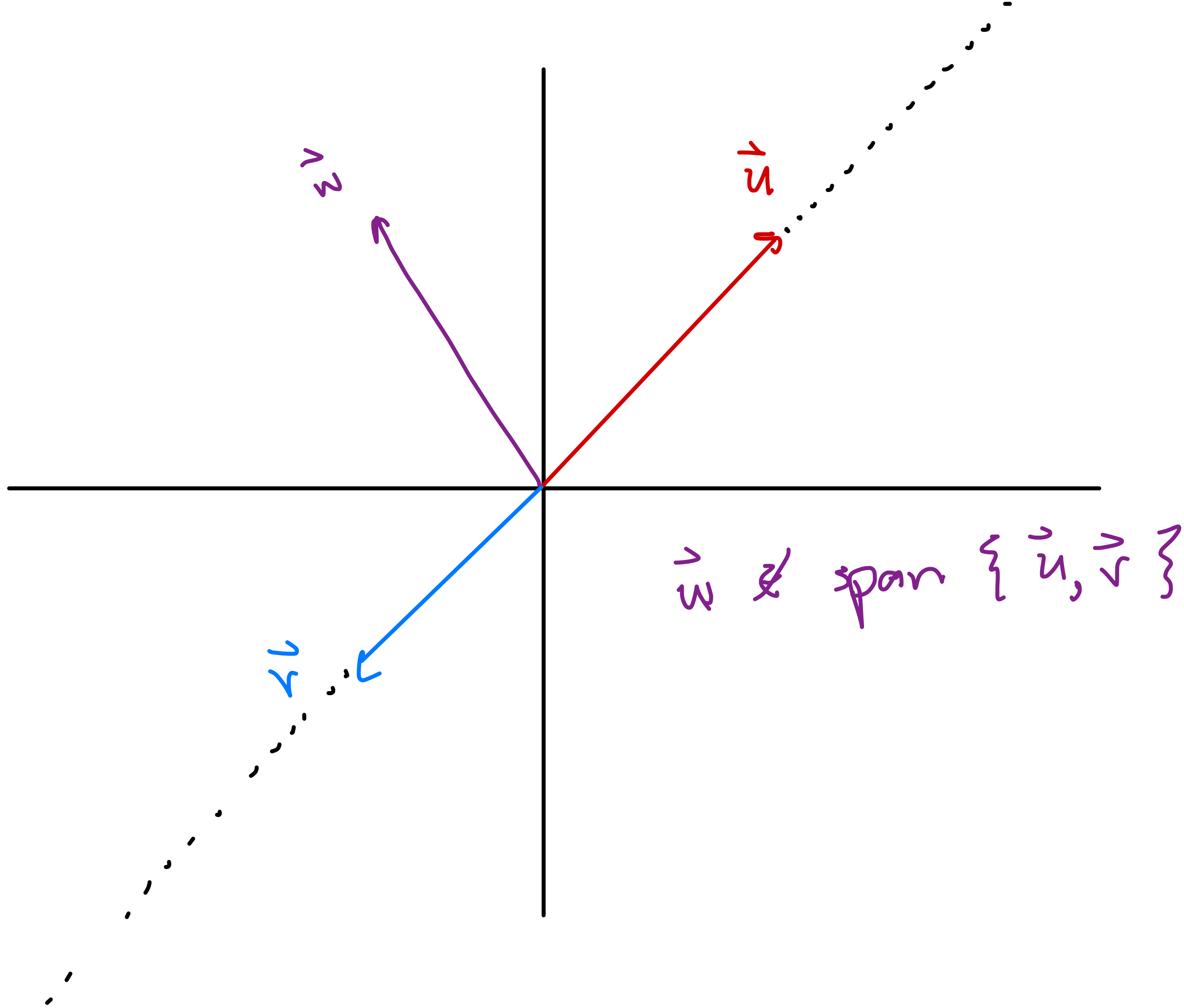
# Spanning all of $\mathbb{R}^2$

if two (or more) vectors in  $\mathbb{R}^2$  span a plane,  
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this will happen  
exactly if they are  
both nonzero and don't  
lie on the same line





# What about $\mathbb{R}^n$ ?

When do a set of vectors span all of  $\mathbb{R}^n$ ?

When do a set of vectors "fill up"  $\mathbb{R}^n$ ?

# A Few Questions

Can two vectors in  $\mathbb{R}^3$  span all of  $\mathbb{R}^3$ ?

Is it required that five vectors  $\mathbb{R}^3$  span all of  $\mathbb{R}^3$ ?

# A Thought Experiment

suppose I give you the augmented matrix of a linear system but I cover up the last column

$$\left[ \begin{array}{ccc|c} 1 & 2 & 3 & \blacksquare \\ 2 & 1 & 0 & \blacksquare \end{array} \right]$$

# A Thought Experiment

then we reduce it to echelon form

$$\begin{bmatrix} 1 & 2 & 3 & \blacksquare \\ 2 & 1 & 0 & \blacksquare \end{bmatrix}$$

# A Thought Experiment

then we reduce it to echelon form

$$\left[ \begin{array}{ccc|c} 1 & 2 & 3 & \blacksquare \\ 2 & 1 & 0 & \blacksquare \end{array} \right]$$

$$R_2 \leftarrow R_2 - 2R_1$$

# A Thought Experiment

then we reduce it to echelon form

$$\begin{bmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \end{bmatrix} \begin{array}{c} \blacksquare \\ \blacksquare \end{array}$$

# A Thought Experiment

then we reduce it to echelon form

$$\begin{bmatrix} 1 & 2 & 3 & \blacksquare \\ 0 & -3 & -6 & \blacksquare \end{bmatrix}$$

Does it have a solution?



# A Thought Experiment

then we reduce it to echelon form

$$\begin{bmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \end{bmatrix} \begin{array}{c} \blacksquare \\ \blacksquare \end{array}$$

$\emptyset \quad \emptyset \quad \emptyset \quad \text{shaded}$

Yes. It doesn't have an inconsistent row

# A Thought Experiment

what about this system?

$$\begin{bmatrix} 1 & 1 & 2 \\ 2 & 2 & 4 \end{bmatrix} \begin{bmatrix} \phantom{0} \\ \phantom{0} \end{bmatrix}$$

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what about this system?

$$\begin{bmatrix} 1 & 1 & 2 \\ 2 & 2 & 4 \end{bmatrix} \begin{bmatrix} \phantom{0} \\ \phantom{0} \end{bmatrix}$$

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# A Thought Experiment

what about this system?

$$\begin{bmatrix} 1 & 1 & 2 & \blacksquare \\ 0 & 0 & 0 & \blacksquare \end{bmatrix}$$

it depends...

# Pivots and Spanning $\mathbb{R}^m$

$$\begin{bmatrix} 1 & 2 & 3 & \blacksquare \\ 2 & 1 & 0 & \blacksquare \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 3 & \blacksquare \\ 0 & -3 & -6 & \blacksquare \end{bmatrix}$$

# Pivots and Spanning $\mathbb{R}^m$

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If it doesn't matter what the last column is,  
then **every choice must be possible**

# Pivots and Spanning $\mathbb{R}^m$

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If it doesn't matter what the last column is,  
then **every choice must be possible**

**Every vector in  $\mathbb{R}^2$  can be written as a linear  
combination of  $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ,  $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$ , and  $\begin{bmatrix} 3 \\ 0 \end{bmatrix}$**



# Spanning $\mathbb{R}^m$

**Theorem.** For any  $m \times n$  matrix, the following are logically equivalent

1. For every  $\mathbf{b}$  in  $\mathbb{R}^m$ ,  $A\mathbf{x} = \mathbf{b}$  has a solution
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# HOW TO: Spanning $\mathbb{R}^m$

**Question.** Does the set of vectors  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$  from  $\mathbb{R}^m$  span all of  $\mathbb{R}^m$ ?

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**Solution.** Reduce  $[\mathbf{a}_1 \ \mathbf{a}_2 \ \dots \ \mathbf{a}_n]$  to echelon form and check if every row has a pivot

**!! We only need the echelon form !!**

# Question

Do  $\begin{bmatrix} 2 \\ 2 \\ 3 \end{bmatrix}$  and  $\begin{bmatrix} 0 \\ 1 \\ 2023 \end{bmatrix}$  span all of  $\mathbb{R}^3$ ?

**Answer: No**

$$\text{max pivots} = \min(m, n)$$

the matrix

$$\begin{bmatrix} 2 & 0 \\ 2 & 1 \\ 3 & 2023 \end{bmatrix}$$

cannot have more than 2 pivot positions

**Not spanning  $\mathbb{R}^m$**

$$\begin{bmatrix} 1 & 1 & 2 & \blacksquare \\ 2 & 2 & 4 & \blacksquare \end{bmatrix} \sim \begin{bmatrix} 1 & 1 & 2 & \blacksquare \\ 0 & 0 & 0 & \blacksquare \end{bmatrix}$$

**Not spanning**  $\mathbb{R}^m$

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In this case the choice matters

We can't make the last column  $[0 \ 0 \ 0 \ \blacksquare]$  for nonzero  $\blacksquare$

But we can make the last column parameters to find equations that must hold

**Not spanning**  $\mathbb{R}^m$

$$\begin{bmatrix} 1 & 1 & 2 & b_1 \\ 2 & 2 & 4 & b_2 \end{bmatrix}$$

$\sim$

$$\begin{bmatrix} 1 & 1 & 2 & b_1 \\ 0 & 0 & 0 & b_2 - 2b_1 \end{bmatrix}$$

$\rightarrow$   
 $R_2 \leftarrow R_2 - 2R_1$

*echelon form*

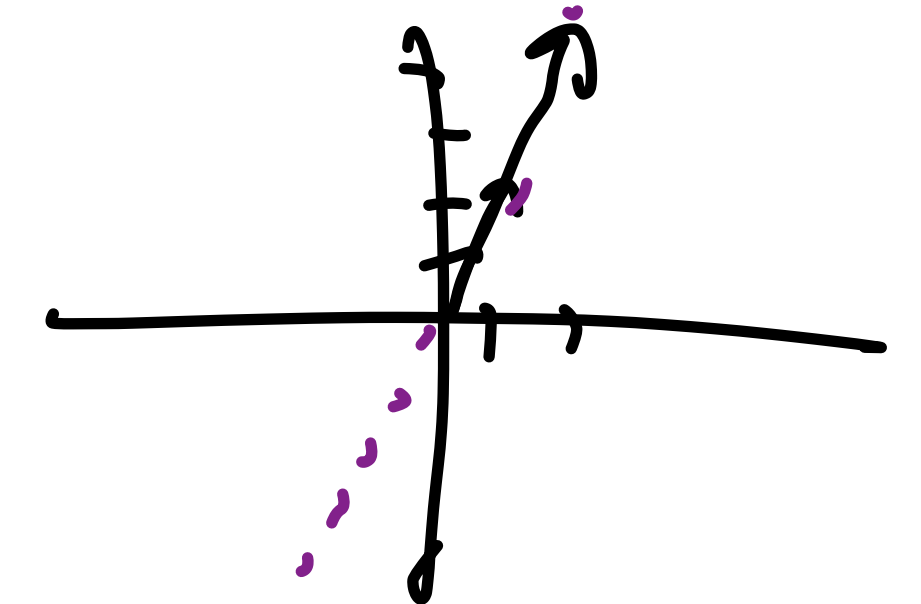
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As long as  $(-2)b_1 + b_2 = 0$ , the system is consistent

**Not spanning**  $\mathbb{R}^m$

$$A \sim \begin{bmatrix} 1 & 1 & 2 \\ 0 & 0 & 0 \end{bmatrix}$$



$$A = \begin{bmatrix} 1 & 1 & 2 & b_1 \\ 2 & 2 & 4 & b_2 \end{bmatrix} \sim \begin{bmatrix} 1 & 1 & 2 & b_1 \\ 0 & 0 & 0 & b_2 - 2b_1 \end{bmatrix}$$

As long as  $(-2)b_1 + b_2 = 0$ , the system is consistent

This gives use a linear equation which describes the span of  $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$  and  $\begin{bmatrix} 2 \\ 4 \end{bmatrix}$

$$-2x_1 + x_2 = 0$$

$$x_1 = 7$$

$$x_2 = 14$$

$$\begin{bmatrix} 7 \\ 14 \end{bmatrix}$$

$$\in \text{span} \left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \end{bmatrix} \right\}$$

$$= \left\{ \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} : -2x_1 + x_2 = 0 \right\}$$

$$\text{span} \left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \end{bmatrix} \right\} =$$

$$\left\{ \alpha \begin{bmatrix} 1 \\ 2 \end{bmatrix} : \alpha \in \mathbb{R} \right\}$$

$$= \left\{ \begin{bmatrix} \alpha \\ 2\alpha \end{bmatrix} : \alpha \in \mathbb{R} \right\}$$

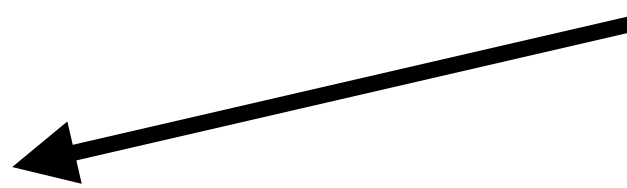
# Question (Understanding Check)

**True or False**, the echelon form of any matrix has at most one row of the form  $[0 \ 0 \ \dots \ 0 \ \blacksquare]$  where  $\blacksquare$  is nonzero.

**Answer: True**

$$\begin{bmatrix} 0 & \blacksquare & * & * & * & * & * & * & * & * \\ 0 & 0 & 0 & \blacksquare & * & * & * & * & * & * \\ 0 & 0 & 0 & 0 & \blacksquare & * & * & * & * & * \\ 0 & 0 & 0 & 0 & 0 & \blacksquare & * & * & * & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \blacksquare \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \blacksquare \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

leading entry not to the right



this is not in echelon form



# Question (More Challenging)

Give a linear equation for the span of the  
vectors  $\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$  and  $\begin{bmatrix} -1 \\ -1 \\ -1 \end{bmatrix}$ .

**Answer**

$$\begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} \quad \begin{bmatrix} -1 \\ -1 \\ -1 \end{bmatrix}$$

# Answer

$$\begin{bmatrix} 1 & -1 & b_1 \\ 2 & -1 & b_2 \\ 0 & -1 & b_3 \end{bmatrix}$$

# Answer

$$\begin{bmatrix} 1 & -1 & b_1 \\ 0 & 2 & b_2 - 2b_1 \\ 0 & -1 & b_3 \end{bmatrix}$$

$$R_2 \leftarrow R_2 - 2R_1$$

# Answer

$$\begin{bmatrix} 1 & -1 & b_1 \\ 0 & 2 & b_2 - 2b_1 \\ 0 & 0 & b_3 + (1/2)(b_2 - 2b_1) \end{bmatrix}$$

$$R_3 \leftarrow R_3 - (1/2)R_2$$

# Answer

$$\begin{bmatrix} 1 & -1 & b_1 \\ 0 & 2 & b_2 - 2b_1 \\ 0 & 0 & b_3 + (1/2)(b_2 - 2b_1) \end{bmatrix}$$

$$R_3 \leftarrow R_3 - (1/2)R_2$$

**Answer**

$$0 = b_3 + (1/2)(b_2 - 2b_1)$$

**Answer**

$$b_1 - (1/2)b_2 - b_3 = 0$$



**Answer**

$$x_1 - (1/2)x_2 - x_3 = 0$$

# Taking Stock

# Four Representations

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} & b_1 \\ a_{21} & a_{22} & \cdots & a_{2n} & b_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} & b_m \end{bmatrix}$$

augmented matrix

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} & b_1 \\ a_{21} & a_{22} & \cdots & a_{2n} & b_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} & b_m \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

matrix equation

$$\begin{aligned} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n &= b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n &= b_2 \\ &\vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n &= b_m \end{aligned}$$

system of linear equations

$$x_1 \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix} + x_2 \begin{bmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix} + \cdots + x_n \begin{bmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

vector equation

# Four Representations

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & b_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} & b_m \end{bmatrix}$$

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$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & b_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} & b_m \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

matrix equation

**they all have the same solution sets**

$$\begin{aligned} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n &= b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n &= b_2 \\ &\vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n &= b_m \end{aligned}$$

system of linear equations

$$x_1 \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix} + x_2 \begin{bmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix} + \dots + x_n \begin{bmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

vector equation

# Summary

Matrix and vectors can be multiplied together to get new vectors

The matrix equation is another representation of systems of linear equations

**Looking forward:** Matrices *transform* vectors